

Computer Science and Computational Thinking in Primary Schools

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Dedicated to my parents,
my grandpa Terry,
and me.

Abstract

As computational devices have become ingrained in modern society, a basic understanding of these, and the Computer Science theories they operate on is now a critical component in understanding our world. Because of this, the subjects Computer Science, programming, and Computational Thinking are becoming a part of school curricula around the world. However, effective integration and teaching of these topics at a pre-tertiary level faces many challenges, as these have not traditionally been taught in school, in the majority of countries. Additionally, many of the claims surrounding how and why these subjects, particularly Computational Thinking, should be taught have not been extensively investigated. The work reported on in this thesis aims to address several of these challenges, with a specific focus on primary school curriculum.

The goals of this thesis are to 1) establish if Computer Science (CS) and Computational Thinking (CT) can be taught effectively in a typical primary school environment; 2) investigate the impact that studying CS and programming can have on CT skills; 3) investigate the impact that developing CT skills can have on general problem solving skills; and 4) document additional impacts, positive or negative, that studying CS, programming, and CT may have on primary school students.

To begin addressing these aims, discussions of four topics are first presented. The many motivations for teaching CS and CT in primary school, and the ambitions of new curricula, are catalogued; a comprehensive literature review on CT, and its place in education, is presented; a history of CS and CT in the New Zealand curriculum is documented, to give context to work conducted with New Zealand schools through this research; and, a literature review on CS and programming teaching approaches and resources is presented.

Three studies, conducted in New Zealand primary schools during the years 2014-2017, are then documented. This began with an exploratory pilot study with one school, then a wide-scale teacher study that expanded on the pilot, followed by a targeted intervention study, focussed on validating the results and conclusions of these prior studies. In total, over 50 teachers and two thousand students participated in these. The initial pilot study and wider teacher study also addressed the first goal of this thesis, how these topics can be taught in primary schools, by providing ongoing evaluations of teaching resources used by participants.

The contributions to knowledge that this thesis makes are: evidence to support the previously un-validated theory that developing CT skills can improve students' problem

solving skills; evidence that, while students who develop CT skills can transfer these to other areas of learning, this requires facilitation and is unlikely to occur implicitly; a map of connections between CS, programming, and CT, and the core skills the NZ curriculum aims to develop in students; evidence that concepts from CS, programming, and CT can be taught effectively at all levels of primary school; observations of disengaged students, who struggle with the usual curriculum, being highly successful and engaged with these subjects; an evaluation of a collection of CS Unplugged resources; and, a review of existing CS school curricula, and a framework CS and CT curriculum that contributed to the design of the new NZ Digital Technologies|Hangarau Matahiko curriculum.

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Chapter I

Introduction

Over the past five years there has been a paradigm shift in school education as Computer Science, programming, and Computational Thinking have entered curricula in New Zealand, and around the world. As computers and digital devices are now ingrained in our everyday lives, basic knowledge of Computer Science, and how these devices function is of immense benefit. Without this understanding, a person is likely to be disadvantaged in our society. To provide their students with this necessary knowledge, education systems around the world are working to introduce Computer Science and Computational Thinking to their schools. However, as these subjects have not previously been taught at school in the majority of countries, there are many difficulties in creating curriculum, training and supporting teachers to deliver it, in evaluating how these subjects can be taught, and even in identifying what we should be teaching. This thesis aims to explore the field of Computer Science and Computational Thinking education, and answer some of these questions.

I began my work in the field of Computer Science Education at what some would consider the ideal time, right as it started to rapidly grow. The literature in this area and the real-world context of my work has evolved constantly, influenced by political and educational developments in New Zealand, and around the world. The state of Computer Science education in 2014, when I first entered this field, differs greatly from what we see today in 2019.

Throughout my research, I have worked on several separate projects that strongly influenced, and were influenced by, my research. At times during 2015-2017, I was involved in the direction and development of the new Digital Technologies Curriculum for New Zealand primary schools, which was introduced to schools in 2018. This work is not fully documented here but was carried out alongside this research. These developments in the NZ and international curricula are discussed several times through this thesis.

Since 2014 I have been a member of the University of Canterbury Computer Science

Education Research Group¹ (CSERG), and this group has had a vital role in my work. The Computer Science Unplugged resources and training workshops developed by our group were used throughout my studies, and the results of these studies have likewise impacted the CSERG's work. This group will be referred to frequently throughout this thesis.

1.1 Research Goals

This thesis aims to overcome some of the challenges to effectively introducing Computer Science, programming, and Computational Thinking to primary school education, and determine whether the integration of these subjects has the desired positive effects on students. The questions I attempt to address in this thesis are listed below. Each of the questions contains several sub-questions, which needed to be answered in addressing the overall question:

RQ1 Can Computer Science and basic programming skills be taught in a typical primary school environment?

RQ1.1 What topics in Computer Science and programming should be trialled for teaching in this research?

RQ1.2 What resources and training can teachers be provided with to successfully teach these?

RQ1.3 Do students learn, and how much do they learn, through these lessons?

RQ2 Can learning Computer Science and programming develop, or improve, Computational Thinking skills for the majority of students?

RQ2.1 How can students' Computational Thinking skills, and changes in these, be measured?

RQ2.2 Do students exercise or employ Computational Thinking skills during Computer Science and programming lessons, based on teacher observations?

RQ2.3 Is there evidence of students' Computational Thinking skills improving, as a result of the Computer Science and programming lessons, based on objective assessment and teacher observations?

¹The CSERG: <https://www.canterbury.ac.nz/engineering/schools/csse/research/cse/>, and <https://www.canterbury.ac.nz/engineering/schools/csse/computer-science-education/>

RQ3 Can learning Computer Science and programming, and developing Computational Thinking skills, have an impact on problem solving skills or general learning, for the majority of students?

RQ3.1 Do students exercise or employ general problem solving skills during Computer Science and programming lessons, based on teacher observations?

RQ3.2 Is there evidence of students' problem solving skills improving, based on teacher observations?

RQ3.3 Is there evidence of the Computer Science and programming lessons having impacts on students' learning in other curriculum areas, based on teacher observations?

RQ3.4 Is there evidence that developing Computational Thinking skills results in these skills being transferred to other areas of students' learning, based on teacher observations?

RQ4 What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students?

1.2 Thesis Contributions

This thesis makes seven key contributions to the fields of Computer Science, programming, and Computational Thinking curricula and teaching:

1. Provides evidence to support the claim that Computational Thinking can benefit the majority students' problem solving skills, in both computing and unrelated fields. This claim had not been tested or validated prior to this research.
2. Provides evidence against the claim that by learning Computational Thinking skills through Computer Science and programming students will automatically transfer this learning to other areas. However, it does provide evidence that the majority of students can transfer these skills *if* they are explicitly taught to apply Computational Thinking skills in different contexts, or if Computer Science and programming are taught in a cross-curricula way.

3. Maps the general skills (the Key Competencies, such as working together, social skills, and resilience) that are the core principles of the New Zealand Curriculum, to Computer Science, programming, and Computational Thinking.
4. Shows that concepts from Computer Science and programming can be taught in a suitable and successful way for all primary school age groups (ages 5 - 13 years old). The majority of prior research has focused on tertiary level education or selected students, whereas the results here apply to typical primary school classroom environments, with generalist teachers; and for students with a wide array of ability levels, subject interest, and difficulties or strengths with learning, with the caveat that teachers are adequately supported and resourced.
5. Shows that not only could these subjects be taught to students with a wide range of abilities, these students were all (apart from students with severe developmental disabilities) able to achieve success to some degree in these classes, and some students who struggled with learning, or were usually seen as low achieving, were particularly successful, often to the surprise of the teacher. Programming and topics from Computer Science appeared to appeal strongly to groups of students who have not been successful within the existing education system, re-engaged them with learning, provided them with new ways to learn and demonstrate their learning which are more suitable for them, and have facilitated the development of other skills, particularly social skills, for these students.
6. Gives an evaluation of multiple Computer Science Unplugged activities, that was done in conjunction with participating teachers. The Computer Science Unplugged resources were found to be easy for teachers to use, the activities were engaging and enjoyable for students, and teachers felt that both they and their students learnt from these. The use of these activities led teachers to continue teaching Computer Science topics after the conclusion of the study; developed their interest in, and knowledge of the subjects; and enthused many to encourage and support other teachers to integrate Computer Science and programming into their teaching.
7. Collates information on the design of Computer Science, programming, and Computational Thinking curricula, and a given a framework for a curriculum, parts of which

have led directly into the design of the new New Zealand Digital Technologies|Hangarau Matahiko curriculum.

1.3 Thesis Structure

This thesis is separated into two parts. Part I, *A New Curriculum for a Digital World*, covers the motivations for this work and functions as a literature review addressing research questions 1.1, 1.2, and 2.1. Part II, *Trialling in New Zealand Primary Schools*, documents three studies carried out in schools between 2014 and 2017, and addresses the remaining research questions.

In Part I, Chapter 2 first outlines the many motivations for adding computing to primary school curricula, and how this work is intended to support this. Chapter 3 is a literature review on Computational Thinking, a concept that has become central to the field of Computer Science education and the design and implementation of school curricula. Chapter 4 catalogues changes in international curricula, as many countries have moved to integrate Computer Science and Computational Thinking into schools; describes the New Zealand curriculum to give context for this research, which took place within the New Zealand curriculum framework; and, gives an overview of the changes in this curriculum over time, as Computer Science and Computational Thinking (under the subject name ‘Digital Technologies’) have been added to this. Chapter 5 is a literature review on Computer Science and programming education and discusses the teaching resources that were used throughout the work covered in Part II of this thesis.

In Part II, three studies, conducted with New Zealand primary school teachers and students, are covered. Chapter 6 first gives an overview of these studies, the methodology used in these, and the resources used for supporting teachers throughout this work. Chapter 7 covers a pilot study conducted with one teacher at an intermediate school, and the implications this had for the following work. Chapter 8 covers a wider scale study, which took place throughout 2015 and 2016, and involved multiple teachers teaching a range of age groups at different schools. Chapter 9 describes the final study, an intervention study that was conducted within three different schools.

Lastly, Chapter 10 gives a summary of the thesis, the contributions it makes, its limitations, and potential directions for future research.

1.4 Terminology

The following terms are frequently used throughout this thesis and their meanings are discussed when relevant. For clarity, summarised definitions are provided here:

Coding: Often used interchangeably with the word ‘programming’, particularly in the popular press and by organisations promoting the ‘learn to code’ movement. It can refer to computer programming, but ‘coding’ and ‘code’ are also used in other areas of Computer Science to mean different things. For the purposes of discussion in this thesis the definition in which ‘coding’ is one part of the act of ‘programming’ will be used. ‘Coding’ can be seen as the final step in the process of programming: taking a designed algorithm/solution to a problem, and implementing this in a form a computer can interpret, i.e. writing it in a programming language. Following from this, ‘code’ can be defined as the actual program, or multiple programs, that have been written [69].

Computational Thinking: Computational Thinking can be described, very broadly, as a collection of mental processes and problem solving methods that are directly related to computational concepts, and the thinking skills and practices used by Computer Scientists. This subject is explored in Chapter 3.

Computer Science: The study of computers, computing concepts, and their interaction with the world. It encompasses the theoretical study of computation and algorithms, and the practical application of this in software and hardware. It covers a range of fields, including but not limited to: computational complexity, algorithms and data structures, artificial intelligence, computer graphics, networks, programming language theory, human-computer interaction, computer architecture, security, and machine learning [42, 59, 136, 182].

Computers and Digital Devices: The general definition of a Computer or Digital Device is a device which processes information, represented as digits (most commonly in binary), by performing mathematical and logical operations, according to instructions given to it, in the form of a program. These terms are frequently used interchangeably, and in the NZ curriculum they are taken to mean the same thing. They will be used in this way throughout this thesis.

Computing: A broader field than CS. It encompasses Computer Science, as well as the links between Computer Science and other disciplines and applications. This includes, for example, the fields of information science and studies, information technology, software engineering, and computer engineering [90].

Digital Literacy: Having the skills and knowledge to effectively use computers and digital devices to store, manipulate, create, and find information [17]. It is focused on learning to *use* digital devices and technologies. It is distinctly different from learning computing, but equally important. It is sometimes used as a synonym for ‘computer literacy’, and ‘digital competence’, but opinions on whether these should be considered the same thing are divided.

Programming: Can be broadly described as the process of taking a problem, analysing it, creating an algorithmic solution to this, and implementing it in a form that a computer can read and execute. The implementation of this generally involves the writing of a program in a programming language and testing the solution. The meaning of programming has evolved over time, and will possibly continue changing as our use and understanding of computational devices changes over time [26, 69].

Part I

A New Curriculum for a Digital World

Chapter II

Motivation

“Everyday life is increasingly regulated by complex technologies that most people neither understand nor believe they can do much to influence.”

Albert Bandura, 2001.

In this digital age, where the influence of computing technology extends to all aspects of society, understanding the nature of computing has become almost as crucial as learning to read and write [10, 161]. Computing is ubiquitous. Our jobs, infrastructure, and education systems depend on it. Its influence has entered national politics and international relations. If someone lacks basic knowledge of computing and digital technologies, they are now placed at a distinct disadvantage to those that do. Yet many people are currently not familiar with these topics.

A core function of school education is preparing students to be capable and informed citizens. To achieve this in the digital age, education systems around the world are changing. The societal impact of computing technologies has motivated many countries to introduce Computer Science (CS) to their school curricula. However, introducing CS is a massive shift in traditional school curricula for the majority of countries. CS and related subjects, such as programming and data science, have not commonly been taught at this level before, and the majority of policymakers and school teachers are unfamiliar with it. This is also the case for New Zealand, where this shift began when the subject area *Digital Technologies* was introduced to high schools in 2011, and primary schools in 2018 [174, 175]. To successfully equip the next generation with a sufficient understanding of computing and digital technologies, we must ensure that any introduced curriculum achieves its educational goals.

In this chapter, I will cover why it is now considered essential for everyone to have a basic knowledge of the nature of computing, and a degree of experience with CS and digital technologies. The reasons for this extend from benefits on individual, national, and broader societal levels, to minimising the negative impacts of this knowledge being restricted to a

small, and relatively homogeneous group of people. Based on these impacts, I then discuss why the inclusion of this subject in school education is important, with a specific focus on primary school. Finally, the motivation for my work, and how it contributes to the success of this new curriculum, is summarised.

The terms “computing”, “programming” and “coding” are used frequently in this chapter; for definitions of how these are used in this context, please refer to section 1.4, page 6.

2.1 Necessary Knowledge in a Digital Society

Humans no longer exist solely in a physical world, but also a digital one. To deeply understand the world we live in, an understanding of CS and the nature of computation has become a necessity. This need was succinctly summarised by Douglas Rushkoff in 2010, in his influential book “Program or be Programmed”. He advocated that programming skills have now become crucial for everyone to learn [152, 153].

“When human beings acquired language, we learned not just how to listen but how to speak. When we gained literacy, we learned not just how to read but how to write, and as we move into an increasingly digital reality, we must learn not just how to use programs but how to make them. In the emerging highly programmed landscape ahead, you will either create the software or you will be the software. It’s really that simple: Program, or be programmed. Choose the former, and you gain access to the control panel of civilization. Choose the latter, and it could be the last real choice you get to make.”

“We teach kids how to use software to write, but not how to write software. This means they have access to the capabilities given to them by others, but not the power to determine the value-creating capabilities of these technologies for themselves.”

- Douglas Rushkoff, 2010, *Program or be Programmed*

The title of this book has become somewhat of a catchphrase in media discussing (and sometimes warning of) the influence of computer technology in our lives.

On the one hand, as these technologies have such a strong impact on our world, those of us working in this field are frequently contributing to solutions for critical problems

facing humanity today [81]. Not only scientific research and knowledge, but also social, economic, and environmental, partially depend on the field of computation. On the other hand, the application of these technologies can have many negative ramifications, both intentionally or through unintentional consequences. These range from wrongful termination of employees [138], to enforcing racial discrimination [138], and influencing our political decision making without our knowledge [66].

Both these positive and negative cases are strong motivators for changing our education systems. These systems now need to include the subject of CS (which often appears as Computational Thinking in school curricula).

2.1.1 Understanding and influencing our world

School education should ready students to be informed citizens in society. A requirement of this is having an understanding of why the world around them functions the way it does. The following quote from Mark Guzdial, author of “Learner-Centered Design of Computing Education: Research on Computing for Everyone”, summarises the contribution of computing to this understanding:

“Computing is a part of students’ lives. We ask students to study chemistry because they live in a world where there are chemical interactions. We ask students to study biology because they live [in] a living world. They also live in a computational world, and the reality of computation is probably going to impact their daily lives more than remembering the structure of a benzene ring or the stages of mitosis.”

Mark Guzdial, 2015 [90]

I would add that this applies to physics as well; we have students study physics because they live in a physical world, governed by physical laws. By this logic, living in a digital world requires the same level of study is directed towards computing. In a similar vein, the case that computation and CS should be considered the ‘fourth great domain of science’ was put forward by Denning in 2009 [64]. He argued that rather than being a part of other sciences, or used in them, it is an entirely separate science unto itself, as it not only occurs in systems created by humans but also in living and social systems. Because of this, it can be called a science of both the natural and the artificial.

Basic knowledge of the sciences actively contributes to a persons ability to understand and influence the world around them. Enabling the next generation (and preferably older

generations as well) to be creators and influencers of new technologies, not just users of it, is a key motivation for teaching computing [115].

Being equipped with these skills and knowledge can assist the next generation in making positive changes in our world. Having a more diverse group of people working in this field, who understand the ethics of using these technologies, is critical. It could increase the socially beneficial impacts of computing technologies, compared to those impacts which could be seen as damaging [81].

By studying computation students can develop a more complete understanding of the world around them. Without it, they may be missing a crucial piece of knowledge, and be put at a disadvantage in both their careers and everyday lives.

2.1.2 Awareness of technology's influence

Being a part of modern society now comes with the condition we accept that digital technologies are ever-present in our daily lives, have a degree of influence over us, and have a degree of access to our personal information. Big data, Machine Learning, and Predictive algorithms have become commonly discussed issues in the media, particularly with a focus on their potential misuse. The technologies of data science and machine learning algorithms have been applied to solving many pressing and complicated problems. They have applications ranging from increasing the accuracy of cancer diagnosis [192], to predicting extreme weather events due to climate change [124]. However, there are sometimes unexpected or, sadly, intentionally damaging consequences from their use.

The current and future issues surrounding the use of big data and machine learning algorithms were explored by Cathy O’Neil in her 2016 book “Weapons Of Math Destruction: How Big Data Increases Inequality And Threatens Democracy” [138]. O’Neil discussed the consequences of the application of these algorithms, and how these frequently disproportionately impact those already struggling in society. For example, the use of predictive policing software, trained with racially biased data, caused police to target predominantly African-American and Hispanic neighbourhoods [138]. The same software has been used in the UK [137], and similarly flawed software in Australia [119]. Algorithms trained with biased data will perpetuate biases, such as prejudice based on peoples race, gender, class, disability, etc. Without a foundational knowledge of algorithms and software, it is challenging to understand, and critically examine, cases such as these.

Another frequently discussed issue surrounding technology and the internet is privacy.

The collection of personal data to be shared, sold, or used to profile people, is commonplace. It is done by private companies, governments, and public service institutions around the world. This can have many positive effects on peoples lives, but this data can also be used in ways that the people whose information has been collected, do not want it to be used. While this is a highly publicised issue, many people are still unaware of just how much of their personal information can be collected, and how it can be used. By learning about software, and the digital technologies employed by these institutions, people can make more informed decisions about how they share their information. They gain more control over their own data, enabling them to protect it if they wish to.

An awareness of these technologies and their capabilities encourages people to interrogate and be critical of the systems around them, and hold public and private agencies to account for how they use, or misuse, digital technology and data. It allows more people to have a say in the development of these systems, and in how they are ethically used. Linda McIver, the founder and director of the Australian Data Science Education Institute¹, gave the following argument [125] for why children specifically need to be aware of how their data can be collected and used:

“The more our kids understand about data science, the more they’ve learnt about the ethics of it, the implications of it, and how the technology actually works, the better equipped they will be to control its impact on their lives. If we can teach kids to manage, analyse, and communicate data effectively, they will be able to participate in informed conversations about the way data is used, both for and against us”

- Linda McIver, 2018, *We need to arm our kids against the interests of Big Data*

Understanding the underlying principles that digital technologies, data, and algorithms are built on empowers people to become informed citizens in a digital world.

2.1.3 Representation in the computing industry

The technology industry suffers from a lack of diversity in the majority of countries where this industry is established. Computing and software professionals tend to overly represent the more privileged groups in our society. For example, in New Zealand, Māori, Pacifica peoples, and women are all under-represented in computing. According to a 2017 report from the New Zealand Digital Skills Forum, only 36% of students studying CS and IT degrees

¹ Australian Data Science Education Institute <https://adsei.org/>

(undergraduate and postgraduate) identified as female, 8% as Māori, 6% as Pacifica peoples; and women made up less than 30% of IT professionals [131]. In 2015 a report on Māori and ICT stated that while Māori make up 12.5% of the New Zealand (NZ) workforce, they only make up 5.8% of the ICT workforce, and were more likely to be employed in lower-skilled ICT occupations, compared to non-Māori in the same sector [128]. To improve diversity in the technology industry, we need a broad range of people, who represent the community, to be pursuing careers in this field.

A wide range of factors has contributed to the low recruitment and retention rates of women, Māori, and Pacifica peoples in CS and IT. Some of the most common reasons are a lack of prior exposure to these subjects, misconceptions about what CS is, what Computer Scientists and Software Engineers do, and the impact of the many stereotypes surrounding these subjects and whom they are suited to [41, 123].

The stereotype that computing is “just for boys”, and that men are *naturally* more competent and interested in these subjects has historically been widespread [123], despite the evidence against this [24, 25, 156, 162]. According to a much more recent meta-analysis performed by Cai, Fan, and Du (2017) there is evidence that the belief in this stereotype may have lessened over time, but still exists [34]. This stereotype is a self-reinforcing one. This perception that men *naturally* have a higher aptitude for these subjects is skewed dramatically by the gender differences in childhood experience with computing and programming, which again is impacted by the stereotype that computing is an inherently ‘male’ subject [123].

Similar to the stereotype that computing is a male subject, and still currently widespread, there is a perception that computing is a ‘Pākehā² thing’, which has been around since computers first began entering the home in NZ [167]. This pertains to the subject of computing, and also the use of computers.

The term ‘digital divide’ is used in many countries. It refers to the difference in access to computers and the internet between different demographic groups. Many of these divisions are based on a person’s income, race, and whether they live in urban or rural areas. Those without access are significantly disadvantaged in the modern world. In NZ, Māori and Pacifica peoples are significantly less likely than other ethnic groups to have access to a computer or the internet within the home [80, 128]. This further enforces the stereotype that computing is not for these groups of people.

² Pākehā refers to New Zealanders of European descent.

In NZ, digital technologies are used by virtually every person every day. The developers of these technologies should be representative of their users. When they are not, there can be drastic consequences, which are frequently related to disadvantaging the groups of people who are under-represented in the technology industry [123, 138]. Increasing the number of people from these under-represented groups entering IT professions is critical to the health of the technology industry, in NZ and abroad. This is not merely an issue of ‘fairness’ (although this is a motivating factor), but also one of societal and economic benefit.

Increasing diversity in the IT workforce benefits both consumers and businesses. Technology businesses with a gender-balanced workforce (particularly a balance at top-level management), compared to businesses that do not, perform better financially; have teams which have higher productivity, are more likely to meet deadlines, and more likely to stay under budget; and demonstrate greater employee performance [13]. In the specific case of software development, increasing diversity in a team also leads to a higher quality of produced software [6]. The contribution of Māori led technology companies to the economy has risen in recent years, and reached approximately \$93 million in 2017. Māori businesses are also more likely to prioritise and achieve social benefits for their communities.

As a lack of prior exposure to computing topics contributes to the digital divide, and the likelihood of a student pursuing study in this area, it is hoped that introducing CS to primary school education can contribute to an improvement in representation in the tech industry, within NZ and abroad.

2.1.4 Economic considerations

Having a digitally competent population is also necessary on an international scale. It will be one of the most significant contributors to our ability to confront pressing global issues such as climate change, resource depletion, and inequality around the world. It is also crucial on a national scale. In the context of NZ, the health of our economy, our national security, and our ability to tackle national issues depend on this. Digital technologies are impacting political decisions, and changes to our legal system. If only a minority of our population understands the nature of these technologies, there will likely be negative implications, and missed opportunities, for NZ.

There are economic incentives for introducing more young people to CS and programming. Virtually every industry and field of employment makes use of software and benefits from this. The software industry itself is growing and highly paid. Jobs in IT are paid

above the national average in NZ, regardless of the employee’s qualification level (although the average increases with higher qualification levels) [135]. In NZ the median base salary for IT professionals rose by 13% in six months, from August 2017 to February 2018. Four of the top ten highest paid jobs advertised on TradeMe³ in 2017 were in IT and software development [35]. However, NZ is facing a shortage of computing professionals to fill available jobs.

In 2017, the NZ Digital Skills Forum carried out three large scale surveys, to investigate the current and projected demand for digitally skilled workers. They surveyed technology firms, government organisations, and collected profile and recruiting data from LinkedIn. The responses to these represented over 130,000 people employed in IT jobs in NZ. The report they produced concluded there is a strong, and increasing, demand for workers with IT and digital skills [131]. Their findings are summarised in Figure 2.1.

These trends of increasing salaries and job opportunities, along with current and projected skill shortages, are not unique to NZ. According to the same report “Australia is predicting an additional 81,000 ICT jobs by 2022, the United Kingdom forecasts an additional 745,000 digital workers by 2018, Europe expects to be short of 500,000 IT specialists by 2020, and the United States of America estimates there will be over 20 million new technology jobs within two decades”, and that three out of four businesses in the UK reported shortage of digital skills as a challenge for their company [131]. In summary, the reports key finding was there is a great lack of local supply of skilled computing professionals, and “Urgent action needs to be taken, to both increase the number of students studying these subjects and attracting under-represented cohorts”. Primary school curriculum can have a significant impact on this.

2.2 Primary School focus

The main arguments for teaching these topics *specifically* at a primary school level were collated in a paper I co-authored, “A Pilot Computer Science and Programming Course for Primary School Students”, in 2015 [67]. These are:

- Preparing students for future endeavours in computing, in both high school, and post-school [114].

³ TradeMe is a website commonly used for advertising jobs in NZ. Salaries in medical specialities and self-employment are not taken into account in this ranking as these jobs are not advertised on TradeMe.



Figure 2.1: New Zealand's Digital Skills Shortage [131]

- Giving students the confidence and capacity to be more than just *users* of digital devices [115].
- Increased diversity in the computing industry, by reaching students aged 12 or younger, particularly those from unrepresented groups, when they are still forming views of their competence in this subject [123]
- The value of exposing students to the concepts before they reach the demands of

high school, the challenges of adolescence, and the risk of exposure to stereotypes of computing, any one of which may discourage them from learning these concepts [69].

- It is well established that, for the vast majority of people, pre-adolescence is the optimal time for learning new skills [103, 140].
- It can encourage early development of general Computational Thinking skills.

As topics in primary school curriculum are very rarely elective, including CS would allow more students to develop a foundational understanding of computing and digital technologies. This would, of course, be of benefit to any students who continue to pursue this subject, but it would also benefit those who do not. As previously discussed in section 2.1, computing extends to almost all fields of work. Regardless of a person’s career path, knowledge of computing will most likely aid them in this, or at least not place them at a disadvantage. The majority of people are currently constrained to merely using technology, rather than modifying it or creating new technologies. When people see problems or ways they can improve things in the world around them, they will have more tools at their disposal to do something about this if they have gained an understanding of computing.

As discussed in section 2.1, for those who might enjoy studying computing further, primary school curriculum provides an effective way of encouraging them to pursue degrees, training, and careers in this field. This is because exposure to these subjects in school allows students to ‘try them out’, and find out if it is something they are interested in pursuing. A student who does not know what CS is, or has never tried programming, is less likely to pursue tertiary education or employment in this area than a student who does. They are more unlikely to even pursue it at a high school level.

When it comes to exposing and interesting a more diverse range of students in this subject, there is evidence that introducing these topics after primary school is likely to be too late. This is because stereotypical views of the ‘type of people’ this subject is for, may have already set in. A 2011 study of 247 American children, found that by the age of six both girls and boys already displayed, implicitly and explicitly, a belief in the stereotype that ‘maths is for boys, and not for girls’ [52]. While there is evidence that these stereotypes are becoming less common, they do unfortunately still exist [34]. At primary school age, particularly before the onset of puberty and the social and self-image issues which come with this, students are likely to be more open to learning these subjects.

Primary school age is also likely an ideal time for students to learn and develop Computational Thinking (CT) abilities. For the majority of people this time is when their neural-plasticity is at its highest and learning new skills, and retaining these into adulthood, comes more easily [103, 140]. It is also a particularly interesting time for learning CT, and about computing, as one of the core skills for these is logical thinking. In Piaget’s model of child development, the ages identified as when logical thinking is developing were between 7-11 [142]. While these age boundaries have been found to be relatively variable between different people [130], primary school ages are still generally the time when this development happens. Papert noted the potential link between children’s increased ability to learn natural languages, and their ability to learn programming languages: “Programming a computer means nothing more or less than communicating to it in a language that it and the human user can both understand. And learning languages is one of the things children do best” [139].

2.3 Motivation for this work

CS is an entirely new subject for pre-tertiary education in the majority of countries who have added it to their school curriculum. If the goals of teaching CS and CT are to be achieved, there are several problems that need to be solved.

The specific topics and concepts that should be included in this curriculum need to be identified. In choosing these, we must consider what is necessary for *everyone* to learn, and what concepts students should be exposed to so that they can make an informed decision about whether they wish to continue study in this area.

Effective ways of teaching these topics need to be identified. If teachers are not adequately supported in implementing this curriculum, there is no hope of it succeeding [186]. If through lack of effective pedagogy these topics are taught badly it will likely do more damage than if they were not taught at all. Alongside good pedagogy, sufficient and effective teacher training and support methods must be established.

Finally, if CS and CT are effectively taught, we need to assess whether they actually have the desired positive impacts and if there are negative impacts.

In order to address these problems, I conducted four years of studies with NZ primary schools. I worked to support NZ primary teachers in trialling CS, programming, and CT in their classrooms and collected information on theirs, and their students’, experiences. During this time I was also involved with the development of the NZ schools Digital Tech-

nologies curriculum, particularly with the newly added ‘Computational Thinking’ strand. The findings of my work contributed to this development.

Chapter III

Computational Thinking, the 21st Century Skill

Coding has frequently been proclaimed to be ‘the new literacy for the 21st century’, with programming being seen as a skill every student will need in the future. The popularity of ‘coding for all’, and ‘Learn to Code’ initiatives has spread across the globe, and sparked millions of students’ and teachers’ interest in this area. However, recently there has been a shift in focus from merely coding, to *Computational Thinking* (CT) being the crucial skill for all to learn [87, 189].

CT has been added as a core learning area to curricula around the world, including in NZ. Teaching CT is however not a straightforward task. Unlike long-standing school subjects, such as Maths, History, or the Sciences, this is a new and ill-defined area, and there is much debate around the meaning of CT. As of 2014, when I began my research, the view that CT is a crucial skill for all students to learn had been widely adopted by the Computing Education community [85]. It was not a focus of the pilot study I conducted, but subsequently became a central focus of my research, my work with teachers, and my involvement in the development of the NZ primary school Digital Technologies curriculum.

There were many developments in the area of CT throughout the years I conducted the research covered in this thesis. Not all sections of the literature covered in this chapter were available during each study, and a significant amount has been published since these studies were completed [62, 63, 86, 101, 110, 144, 146, 161].

In this chapter, I will cover the history of CT; the many descriptions and definitions of it that have been given, and the challenges of settling on one overarching definition. The relationships between Computer Science, problem solving, and CT, and the implication of these factors on defining CT are discussed. Assessment methods (and the lack thereof) are then explored. Finally, CT frameworks specifically designed for use in primary school education are discussed.

The terms “computing”, “programming” and “coding” are used frequently in this chapter; for definitions of how these are used in this context, please refer to section 1.4, page

3.1 What is Computational Thinking?

Computational Thinking can be described, very broadly, as a collection of mental processes and problem solving methods that are directly related to computational concepts, and the thinking skills and practices used by Computer Scientists. CT is a concept that has existed, in varying forms and with different names since the 1940s, and has recently become popular in the K-12 Education sphere [63, 169].

The term ‘Computational Thinking’ was first introduced in 1980, by Computer Scientist and Educator, Seymour Papert, in his pioneering book “Mindstorms. Children, Computers, and Powerful Ideas” [139]. At that time computers and computer-aided-instruction were becoming common in schools. They were used mainly with ‘drilling and practice’ programs, designed to reinforce previously learnt knowledge [129]. Mindstorms was about very different ways computers could be used in education. It presented the groundbreaking idea that computers could be used as a tool to facilitate new ways to learn, and think:

“In this book I discuss ways in which the computer presence could contribute to mental processes not only instrumentally but in more essential, conceptual ways, influencing how people think even when they are far removed from physical contact with a computer.”

- Seymour Papert. *Mindstorms* (Page 4)

While Papert coined the term Computational Thinking, and *Mindstorms* explored the foundational concepts of it, it was not actually identified as a specific term, or name, for anything. It was simply mentioned once in passing when referring to computer hobbyist clubs: “Their visions of how to integrate computational thinking into everyday life was insufficiently developed” (page 182 [139]). The assignment of this title to the thinking, learning, and problem solving that was deeply explored by Papert, seems to have come from future researchers in this area. Papert’s work was highly influential on the use of computers as a teaching tool in the classroom, but CT as a distinct concept did not truly gain traction in education and computing research until much later.

This happened in 2006 when the idea of ‘Computational Thinking for all’ was reintroduced to the Computing community in Jeannette Wing’s seminal article ‘Computational Thinking’ [189]. She identified it as a “Fundamental skill... something every human must

know to function in modern society”, and has since referred to it as “the new literacy of the 21st century” [191].

Wing’s article triggered a wave of interest in the topic, particularly in the education field. Wing described CT in terms of the many different skills it encompasses, the elements of CS that are key to it, and how computation itself can be applied to almost any field of study. In 2008 Wing published a follow-up article, “Computational thinking and thinking about computers”. This article gave one of the first specific definitions of CT:

“Computational thinking is taking an approach to solving problems, designing systems and understanding human behaviour that draws on concepts fundamental to computing” [190]

Expanding on her 2006 article, she explored the ubiquity of CT and the types of problems it can be applied to, from optimising the assignment of organ donors to recipients, to creating new techniques for analysing historical texts. In 2010 Wing, Cuny, and Snyder published another definition:

“Computational Thinking is the thought processes involved in formulating problems and their solutions so that the solutions are represented in a form that can be effectively carried out by an information-processing agent.” [191]

Wing stated that this definition was inspired by a conversation held with Alfred V. Aho, who then proposed a similar definition in 2011. This definition conceptually agreed with Wing, Cuny, and Snyder’s, but specified that to be carried out by an information-processing agent a solution must be “represented as computational steps and algorithms” [1]. Both of these definitions touch on an essential point about *computation*, and algorithmic solutions. These are not something performed only by computers, they can be performed by any information-processing agent, which includes humans. As CS is the study of computation, being a Computational Thinker is a key part of being a Computer Scientist, and CT is sometimes described (in a slightly over-simplified way) as “thinking like a Computer Scientist”.

Many other definitions have been proposed since these, and over the past five years this area has become a particularly popular topic of research.

3.1.1 *Challenges for finding a definition*

Articulating a broad overarching definition of CT has proved difficult. This is partially because it is generally agreed CT encompasses many varied aspects across different levels of education, and in areas outside of education. Opinions on its importance and breadth vary among CS educators and researchers. With new curricula being introduced across the world, and the high public engagement created by the “Learn to Code” movement, work in this area is additionally subject to political motivations.

The many definitions produced since 2006 vary significantly in form, from one-sentence summaries to comprehensive explanations of each skill encompassed by CT, and their applications. The contents of these definitions also diverge in many ways [169]. Some place a strong emphasis on the connection between CS concepts (and in some cases programming) and CT skills, while others aim to remove the connection to CS entirely. More recent definitions have frequently extended CT to cover attitudes and dispositions towards problem solving, and interpersonal skills. Many have also been crafted for specific age levels and target audiences, such as generalist teachers, contributing to the wide variation in definitions [63, 85, 158, 159].

Wing’s article gave a broad, big picture description of CT, and the many benefits she believed it could have. As it was not intended to outline a CT framework or curriculum it was, unsurprisingly, not particularly suitable for application in education [15, 121]. Many of the definitions published since 2006 have been criticised for being too broad and vague, while other more verbose and detailed ones have been dismissed as too difficult to interpret, particularly for people without a background in Computer Science (CS). The CS education community has not yet agreed on a definition of CT, despite the large number that have been published, and the many similarities between these [122].

Voogt et al, [185] examined the tension of whether to define CT in terms of what are considered its ‘core qualities’, or in terms of the more ‘peripheral qualities’, such as the attitudes and dispositions the skill ideally encourages. They argue that incorporating more than core concepts and skills “runs the risk of diluting the idea of CT, blurring and making it indistinct from other 21st century skills (e.g., media information literacy)” [185]. Tedre and Denning share this view. They claim the recent efforts to broaden CT will have a significant negative impact on the potential educational merits of teaching this topic [169].

Both within, and outside the field of education, no one definition has been agreed upon. This can be seen from the large amount of work that has been done since 2006 to produce

a universal definition of CT [14, 15, 16, 29, 60, 85, 88, 158, 159, 161]. It has been defined and redefined by researchers from universities, industry, international computing organisations, and educational institutions around the world, and is still continuously developing. Meanwhile, curriculum is being explored and developed without an agreed upon definition to build on. This is causing huge variation between curriculum levels around the world. This is discussed in Chapter 4, in relation to synthesising a curriculum for NZ.

Despite the differences in these definitions and descriptions, there are many areas of overlap between them and there is a growing consensus on the necessary elements of a definition [50]. There is also a high degree of consistency with the original thought and problem solving processes described by Papert. By examining the broad range of these definitions it is possible to get a general, and usable, view of CT even if there is no one universally agreed upon definition.

3.2 *Reviews of the definitions*

There have been several attempts in recent years to collate the many CT definitions and frameworks that have been proposed. In 2013, the most prominent definitions of CT were aggregated by Selby and Woolard [159] in a technical report, covering descriptions of CT published between 2006 (beginning with Wing’s 2006 article) and 2012. They examined published documents on CT, papers referencing Wings’ publications, and CS curricula (current and proposed at the time) from six countries. From this review, they identified common terms and concepts and advised which terms should, and should not, be included in a definition. Their conclusions are shown in Table 3.1.

Selby and Woolard’s conclusions broadly match the definitions and descriptions of CT published since their review. The main exception to this is that Selby and Woolard exclude ‘Logical Thinking’ and ‘problem solving’, while these are included in many of the more recent definitions. CS content is also often incorporated into a definition, or its inclusion is suggested when CT is taught.

In 2013 Grover and Pea [85] reviewed papers from the same period (2006-2012) and focused on the importance, and possibilities, of introducing CT to a curriculum. They emphasised that along with finding a suitable definition, assessment of CT is critical to its success in any K-12 curriculum [85]. Despite this, CT assessment is an underdeveloped area. Research on this, which goes beyond small exploratory studies, has only recently been emerging.

Term	Status	Justification
A thought process	Include	Consensus found in the literature
Abstraction	Include	Consensus found in the literature
Decomposition	Include	Consensus found in the literature
Logical thinking	Exclude	Broad term, not-well defined
Algorithmic thinking	Include	Well-defined across multiple disciplines
problem solving	Exclude	Broad term, evidences the use of skills; develops acquisition of skills
Evaluation	Include	Well-defined across multiple disciplines
Generalization	Include	Well defined concept, although the term may not be familiar
Systems design	Exclude	Evidences the use of skills
Automation	Exclude	Evidences the use of skills
Computer Science content	Exclude	Evidences the use of skills
Modelling and simulation	Exclude	Evidences the use of skills

Table 3.1: Computational Thinking Definition Terminology - table by Selby and Woolard, reproduced from [159]

In 2017 Shute, Sun, and Clarke conducted a research review of CT definitions across different disciplines [161]. From this, they presented a definition of CT for K-12 educators and a model for implementing and assessing CT in K-12. They emphasise the distinction between programming and both CS and CT, and that there is much more to CS and CT than programming. This is an important distinction to make, as programming is sometimes seen as the core of both of these areas, which can lead to education focusing solely on programming. This is problematic, as there are many additional concepts and practices encompassed by CS and CT. Webb et al. [186] described the difference, and relationship between programming and CT as follows:

“The distinction between computational thinking and programming is subtle; in principle computational thinking does not require programming at all, although in practice, representing a solution to a problem as a program provides a perfect way to evaluate the solution, as the computer will execute the instructions to the letter, forcing the student to refine their solution so that it is very precise.”

- Webb et al. 2016. *Computer Science in K-12 school curricula of the 21st Century: Why, what and when?*

Shute, Sun, and Clarke, also discuss the differences between CT and other thinking processes, and what makes CT distinct from Mathematical, Engineering, Design, and Systems thinking. However, their proposed definition seems to do the opposite of this, as it is exceedingly general. They define CT as “The conceptual foundation required to solve problems effectively and efficiently (i.e., algorithmically, with or without the assistance of computers) with solutions that are reusable in different contexts” [161]. The qualifiers ‘effectively and efficiently’ and ‘reusable in different contexts’ can be said to encompass CT, but these are in no way unique to CT or algorithmic solutions. This can effectively apply to solutions resulting from any of Mathematical, Engineering, Design, and Systems thinking, and other fields. This definition discounts one of the core ideas that is included in the majority of prominent CT definitions: that the algorithmic solution is one which is made up of *computational steps*, meaning they are specific and unambiguous enough that they can be carried out by an information processing agent.

In 2018, Juškevičienė and Dagienė compiled a timeline of the most prominent definitions and frameworks, for CT, published between 1996 and 2017 [104]. Based on these definitions they selected eight concept groups to describe CT: Data analysis and representation, Computing Artefacts, Decomposition, Abstraction, Algorithms, Communication and

Collaboration, Computing and Society, and Evaluation. *Computing Artefacts* was a new inclusion, and it is generally associated with digital literacy and the *use* of Digital Technologies, rather than CT or CS. They also examined the links between CT and ‘Digital Competence’, as defined by the ‘Digital Competence (DC) Framework 2.0 (DigCom)’.¹ They concluded there was a great deal of overlap between these two areas, which is likely why they included Computing Artefacts in the CT concepts. The concepts included in this particular model of Digital Competence appear to include more skills focused on problem solving than those that are often included in similar models, or interpretations. In contrast, in NZ ‘Digital Competence’ is generally seen as using computers and creating things with them, rather than being focused on problem solving or an understanding of how computers work.

Alan Bundy described the difference between what is generally known as digital literacy or competence, and CT in the following way:

“Of course, we all have computers on our desks nowadays. We all use them for email, web browsing, word processing, game playing, etc. But the computational thinking revolution goes much deeper than that; it is changing the way we think. Computational concepts provide a new language for describing hypotheses and theories. Computers provide an extension to our cognitive faculties. If you want to understand the 21st Century then you must first understand computation.”

- Bundy. 2007. *Computational Thinking is Pervasive* [33]

3.2.1 *The role of Computer Science in Computational Thinking*

For the purposes of this thesis Computer Science is defined as the following:

Computer Science: The study of computers, computing concepts, and their interaction with the world. It encompasses the theoretical study of computation and algorithms, and the practical application of this in software and hardware. It covers a range of fields, including but not limited to: computational complexity, algorithms and data structures, artificial intelligence, computer graphics, networks, programming language theory, human-computer interaction, computer architecture, security, and machine learning [136, 42, 59, 182].

¹ <https://ec.europa.eu/jrc/en/digcomp/digital-competence-framework>

Despite Wing’s assertion that CT draws on concepts from CS [190], there have been arguments that CT should be considered, and therefore defined, as entirely separate from CS and computing [61]. The fact that the computational steps and algorithmic solutions can be performed not only by computers, but also by humans, has been taken by some to mean that CT need not involve programming or computers at all. This viewpoint has become increasingly common since the current CT in K-12 movement began. Many of the definitions published in the past ten years have focused less on computing and CS, and in some cases these topics have been all but removed. This is largely based on the view that CT should be seen as a form of problem solving that is widely applicable, and not unique to computing subjects. Much of the emphasis on CT education is on incorporating it into other subjects and aspects of life, in which case removing its connection with CS could be seen as a logical decision.

On the other hand, it has also been argued that removing the CS elements of it would change the core meaning of CT, and make it indistinguishable from general problem solving skills [169, 185]. An example of this is the previously discussed definition proposed by Shute et al. [161]. This effort to broaden CT, so that it can cover all possible areas, may have also contributed to the difficulty of defining it clearly [63]. CT is a concept that came directly from CS and Computer Scientists. In the 1950’s-80’s the idea that there were thinking and problem solving skills which were unique to CS was discussed almost exclusively in the context of it being one of the contributing factors to what made computing a subject in and of itself, rather than merely a field of engineering or mathematics [169]. According to Tedre and Denning, one of the first people known to have touched on this idea was Alan Perlis, who, in 1960, presented the idea that “the value of computers is less about their use as instruments and more about their cultivating a certain style of reasoning about problems and designing solutions” [169]. The idea that there is a set of thinking skills that are unique to computing has been reiterated by many highly influential figures in the field of CS, such as Dijkstra and Knuth [169].

CT is frequently described using terms that are widely applicable and have no reference to CS or computing, such as ‘thinking process’ and ‘problem solving method’. Removing the computing element of CT effectively removes the attributes that distinguish it from other already established problem solving methods. CT is also described as a process that produces solutions that can be implemented using computers, computing tools, information processing agents, or that in some way leverages computational concepts, to improve them [14, 43, 60, 82, 88, 100, 190]. Enabling people to take advantage of computing technology in the ‘digital

age’ is one of the most referenced benefits of CT education. Removing CS from a definition of CT would therefore be detrimental, as it would make much of the effort to teach the skill redundant [14, 43, 60, 82, 88, 100, 190].

While it is important that a definition of CT should allow it to be applied across the broader curriculum, I argue that divorcing CT from CS will not assist with this, and agree with the view that this would decrease the benefits of CT education. CT is a relatively abstract notion, that can be made more tangible by the CS concepts it draws so heavily from. The inclusion of CS principles would not automatically prevent cross-curricula integration of CT. As will be expanded on in chapters 7 and 8 in Part II of this thesis, during my work with many NZ primary schools it was demonstrated that CS and programming principles could be taught in conjunction with a range of subjects, including literacy, numeracy, geometry, algebra, design, and social studies [67]. During this work, teachers found that using CS and programming concepts strengthened learning, and was highly engaging for students. If CS can be integrated with other subjects, there is no reason why CT, together with CS, cannot be.

It is also important to consider the place of programming in CT education, and as a part of CT in general. The misconception that programming is all there is to CS and computing is widespread, and this sometimes influences the way CT is presented and described [69]. Tedre and Denning describe how “CT initiatives that focus solely on programming tools and techniques market a tasteless, scentless view of computing that emphasises [an] analytical abstract world far distant from the hands-on dirty complexities of the real-world” [169]. There is much more to CS and CT than programming, so it is sometimes argued that programming should be cut out entirely from CT. However, like CS, programming can be used to facilitate and strengthen CT education. When solving problems through CT, the goal is to create an algorithmic solution, made up of computational steps.

Programming provides a way for students to implement these algorithmic solutions. As computers will obey the instructions they are given exactly as they are, programming allows students to check if their algorithmic solutions work as they intended, and enforces the need for these algorithmic steps to be precise and unambiguous. “Programming puts the wheels on CT” has become a way we describe this relationship between CT and programming, in the UC Computer Science Education Research Group.

3.3 Problem solving and Bold Claims

Many claims have been made about how CT skills can benefit problem solving in multiple subject areas, and in everyday life. The problem solving approach of CT is seen by much of the CS education community as one that students should be able to apply to multiple areas, and can assist their general learning.

“The Computer Science for All education movement, which began around 2006, is motivated by two premises: that computational thinking will better prepare every child for living in an increasingly digitalized world, and that computational thinkers will be superior problem solvers in all fields.”

- Peter J. Denning, 2017. *Remaining trouble spots with computational thinking*

(It should be noted that Denning is stating this rather than claiming or agreeing with it, and in context he is critiquing this claim). Despite how widespread this opinion is, there is very little empirical evidence that this learning transfer occurs, and that CT can benefit people other than those who work in, or study, computing [63, 95, 169]. Many of these claims appear in blog posts, opinion pieces, and other ‘grey literature’. Where they appear in research publications, there is little to no evidence given to back these statements. In 2015, Mark Guzdial stated that after reviewing all available research on the benefits of CT for generalised problem solving, there was no adequate evidence to confirm these claims [90]. Several Educational Psychologists have also expressed misgivings about these claims [89]. Much educational research on the topic of learning transfer concludes that “Transfer does not happen easily or automatically” [3], and needs to be facilitated through appropriate ‘teach for transfer’ pedagogy. CT seems to currently be seen as an exception to this.

The claim that CT can benefit *everyone* is seen by many as one of the foundations the CT education movement is based on. However, this overarching benefit is not universally accepted. Denning challenged this assertion that CT will benefit *everyone*, in both their careers and their everyday lives:

“it is reasonable to question whether computational thinking is of immediate use for professionals who do not design computations—for example, physicians, surgeons, psychologists, architects, artists, lawyers, ethicists, realtors, and more.”

“Another claim suggested in the operational definitions [of CT] is that computational thinking will help people perform everyday procedural tasks better—for example, packing a knap-sack, caching needed items close by, or sorting a list of customers. There is no evidence to support this claim. Being a skilled performer of actions that could be computational does not necessarily make you a computational thinker and vice versa.”

- Peter J. Denning, 2017. *Remaining trouble spots with computational thinking*

As these claims are frequently used as a key selling point when advocating for school curriculum changes, establishing their validity is critical to the development of new CS and CT school curricula. This is not to say that CT is not invaluable if these claims turn out to be false, it could still be highly beneficial to students, but if these assumptions are invalid then perpetuating them can be damaging [63, 95, 169].

“Exaggerated claims about CT serve no one—for eventually, when CT cannot deliver on such claims, there will be many disillusioned educators and consumers of education, and computer science will be seen as an over-seller of CT. CT is powerful enough without exaggerated claims.”

- Tedre and Denning. 2016. *The long quest for computational thinking*.

Whether or not teachers observe improvements to students’ general problem solving is a topic I choose to focus on during my work, due to these contentions and the lack of supporting research.

3.4 Assessment

For CS and CT curricula to succeed, educators need to be provided with methods to measure students’ learning [150]. Without a means of assessment, it is difficult for teachers to gauge their students’ understanding of the topics they have covered, determine if the teaching materials used are suitable, and give students feedback on their progress. As CS is not a subject that has traditionally been taught at this level and CT is new to curricula in general, there are currently no widely established methods of testing these in primary school [164, 171, 193]. However, this field is rapidly growing so this may change in the near future.

Many of the assessment methods that have been used or proposed (as of writing) for pre-tertiary education focus on close observation of students, or in-depth analysis of the

work they produce to evaluate their skills [84, 115, 116, 181, 187]. These methods are too labour intensive, and require too much subject matter expertise, to be used for assessing large numbers of students in typical school environments. Thus, they are not currently of practical use in the majority of schools. Interviews with students can provide a measure of their level of CT, and are a powerful method of gaining insight into students learning [78]. However, this is again a time intensive approach, and not suitable for assessing large numbers of students.

In 2014 Curzon et al. published a framework for CT in school education [51], which included a section on assessment with a different approach. This framework described a set of core CT concepts (based on Selby and Woolard’s 2013 review [159]) and provided a set of classroom techniques that could be used to teach each of these concepts. Their assessment framework was composed of a progression of learning outcomes for each CT concept. For example, the first learning outcomes for the topic Algorithms include “Understands what an algorithm is and is able to express simple linear algorithms symbolically. Understands that computers need to be given precise instructions”, and the final set that students are expected to progress too includes being able to design recursive algorithms, and “Understanding that some problems cannot be solved computationally”. Learning outcomes like these are necessary for the development of assessment methods, but on their own are unlikely to be sufficient. They would assist teachers in understanding what their students should be learning. However, without an in-depth knowledge of the subject matter it would be difficult for a generalist teacher to ascertain whether a student had achieved these outcomes, without additional support.

An alternative method of assessing students is through standardised tests. This would not provide the same level of insight into a students CT skills as the previous methods described could. These could also be less accurate if test questions were not well designed, or their design disadvantaged specific groups of students, for example those with lower reading comprehension skills. However, this approach is much more scalable, as it would allow large numbers of students to be assessed in a consistent and objective (assuming unbiased question design) manner. It would remove the risk of an assessors level of understanding impacting their ability to accurately, and objectively, assess students. Removing the need for teachers to have extensive knowledge of this subject area is highly beneficial, while this topic is not widely integrated into pre-service teacher education.

In 2015 Romàn-González proposed a set set of design guidelines for creating questions for CT assessment, and an initial set of 28 questions which had been reviewed by a set of 20

experts in the field of CS education [149]. These questions were used for a CT test, targetted at ages 12-13 years old, that was taken by 1,251 Spanish students between ages 10-16 years old. These students also completed two standardised tests measuring cognitive abilities and problem solving skills, and scores were compared between these three tests. The results of this were reported on by Romàn-González, Pèrez-González, and Jimènez-Fernàndez in 2016, and they found positive correlations between the CT test scores and problem solving test scores, and between the CT test and cognitive abilities results [151]. These results lend support to the validity of the test used, if CT is considered to be linked with cognitive ability and problem solving. As the questions used were thoroughly reviewed by a set of experts before the test was administered, this test may be a robust method of assessing students CT abilities. However, the set of questions is entirely focussed on the use of directional instructions and programming concepts, for example loops and conditional statements. If we accept the current general view of CT, then CT is considered to involve more than programming skills and concepts. While this proposed assessment provides many questions that can effectively evaluate a set of CT skills, its currently programming centric nature limits its effectiveness. This limitation was acknowledged by Romàn-González, Moreno-Leòn, and Robles in 2017[150], and they recommended the use of complementary tools for assessment, specifically Dr Scratch and tasks from the Bebras Challenge.

The previously mentioned *Bebras Challenge* is an extra-curricula CS challenge (also referred to as an Informatics challenge). Challenges and competitions such as this can offer an alternative assessment method, though they also have their own drawbacks due to their limited validation in the context of CT curriculum. The Bebras Challenge was one of the assessment tools used in my research and so is discussed in more detail below.

3.4.1 The Bebras Challenge

The ‘Bebras International Challenge on Informatics and Computational Thinking’² is a popular Informatics challenge, which is run annually [57]. It began in Lithuania in 2004 and has since spread around the world. The founder of the challenge and the organisation that runs it, Prof. Valentina Dagienė, aimed to use the challenge to introduce school students to the subject of Informatics (in this usage, synonymous with CS and computing). The challenge has now been held in over 60 countries, including NZ. Since its inception, the goals of the challenge have expanded to include promoting and developing CT skills, along

² <https://www.bebas.org>

with promoting Informatics, CS, and computing. I choose to use the Bebras Challenge for CT assessment in two of my studies, in the absence of validated and widely established assessment methods.

Much work has been done on evaluating the tasks used in the Bebras Challenge, their relevance to CT, and how they can be used in CS and CT education [12, 23, 38, 53, 54, 56, 57, 97, 98, 102, 184, 194]. For example, Dagienė and Stupuriene [57] mapped Bebras tasks to levels on Bloom’s taxonomy to evaluate which thinking skills the tasks required. This taxonomy describes a hierarchy of six ways a learner can demonstrate their knowledge. These are to: Remember, Understand, Apply, Analyse, Evaluate, and Create. Remembering is the lowest on this hierarchy, and creating is the highest. They found that the majority of Bebras tasks lie at the higher end of Bloom’s taxonomy (Understanding, Applying, Analysing and Evaluating). They also concluded that the tasks encouraged learning experiences for students, a factor that was cited as a requirement for ‘good tasks’ by Dagienė and Futschek [53]. Other criteria which were listed by Dagienė and Futschek were tasks should be independent of any curriculum, particular software or hardware systems, and programming languages. Dagienė and Stupuriene [57] also found that tasks can be broadly divided into five categories: Algorithms and programming; Data, data structures, and representations; Computer architecture and processes; Communications and networking; and HCI. The first two of these are core parts of CT, and the remaining three are all, in some way, related to it. All five of these are core elements of CS.

The broad consensus of this research is that the Bebras contest is an effective way of engaging students in CS/Informatics, and the tasks cover CT concepts. However, it is important to note that this research may be biased as much of it was conducted by individuals who are, or have previously been, involved in running the contest.

There are problems with the challenge that have also been identified. For example, both Bellettini, et al. [23] and Vegt [184] found that the difficulty levels (easy, medium, hard) assigned to different tasks did not accurately reflect the actual difficulty (based on participants results) of the tasks. 37% were found to not agree with the assigned difficulty of the tasks by Bellettini, et al. and between 40% and 72% (across different age groups) by Vegt [184]. The possible impact of reading difficulty on tasks was briefly mentioned by Vegt, and this was a factor I found to be highly influential in my use of the Bebras challenge. Bebras tasks can also be problematic due to the strong focus on creating stories within the questions to try and make them interesting and relevant to learners. This can at times distract from the CS concepts the challenge is meant to test, can be confusing for

learners, and calls in to question whether the test is assessing the intended skills, a problem I discovered when using the challenge in my research. This risk of questions lacking a focus on CS concepts and skills was noted by Romàn-González et al. as a reason for using Bebras alongside other assessment tools, such as their CT test and Dr Scratch, rather than as a single assessment [150].

The suitability of the Bebras challenge as an assessment tool also came in to question in 2017, after I began my final study, when it was used in a randomised controlled trial. This trial ran from June 2015 till September 2016, to assess changes in UK students' CT skills, and compare these changes between students who spent a year taking part in Code Clubs³ and students who did not [165]. This study found that students' scores in the challenge did increase between the pre and post-tests, but there was no difference in this increase between participants who participated in Code Clubs and those who did not. The authors Straw, Bamford, and Styles concluded that "attending Code Club for a year did not have an impact on pupils' computational thinking over and above changes that would have occurred anyway". This was an unexpected result, particularly as the trial also showed students' who attended the Code Clubs improved in coding skills and their teachers reported improvements in their students' problem solving ability. These results and conclusions are very similar to those of the intervention study I conducted, discussed in Chapter 9. Though these results have brought up questions around the validity of the Bebras challenge as an assessment tool, this was not the conclusion of Straw, Bamford, and Styles study [165]. The authors noted that it was possible the Bebras challenge may not have been sensitive enough to detect differences in students' skills, but they were fairly confident, based on their results, that it was. The results of this study were published in 2017, after my final study had begun and I had already selected the Bebras challenge as an assessment tool.

The challenge format and how it is administered is discussed in Chapter 6.

3.5 Operational definition of Computational Thinking for this research

Among the many CT definitions and frameworks that have been proposed are a number that have been created specifically for use at a primary school level. One of the most prominent of these frameworks is the definition given by *Barefoot Computing* [49], the *Computing at School's* primary education working group. In carrying out studies with NZ teachers and schools, it became necessary to select a specific framework to use for analysing the results

³ <https://codeclub.org/en/>

of the studies conducted during 2015, 2016, and 2017 (discussed in chapters 8 and 9). I chose to use an already published framework that had been written and reviewed by a group of subject experts, rather than designing a new one. For this, I selected the Barefoot Computing definition, described in section 3.5.1. As this definition is one of many there are areas where it disagrees with others. For example the inclusion of the CT ‘Approaches’ is in line with the very broad definitions of CT, described by Voogt et al. [185] as those that included the ‘peripheral qualities’, and does not match the definitions that try to stay more focussed on the ‘core qualities’ and skills, that are more closely linked with CS.

The Barefoot Computing definition and framework do not fully agree with my own view of CT, which aligns more closely with Tedre and Denning’s [169], however it was chosen over other definitions as it is specifically targetted at primary school and is more relevant to the NZ curriculum. In particular, there were strong links between this definition and the *Key Competencies* (discussed in Chapter 4) and the contents of the draft NZ Digital Technologies curriculum in 2015 (also discussed in Chapter 4).

3.5.1 Barefoot Computing

Computing at School (CAS) is an organisation, and community, that provides support for K-12 computing education through online resources, and workshops throughout the UK [44].

They use their online resources and forums to provide international support. CAS was a major contributor to the addition of Computing to the English curriculum, and their resources were created to support teachers adopting the England national computing curriculum. These resources are widely applicable to computing education elsewhere, including in NZ. CAS outline their framework for CT in school in “Computational Thinking: a guide for teachers” [49]. This guide uses a combination of Wing’s 2006 and 2011 definitions for CT as a starting point for their framework. They describe CT as follows:

*Computational thinking is a cognitive or thought process involving **logical reasoning** by which problems are solved and artefacts, procedures and systems are better understood. It embraces:*

- *the ability to think **algorithmically***
- *the ability to think in terms of **decomposition***
- *the ability to think in **generalisations**, identifying and making use of **patterns***

- the ability to think in **abstractions**, choosing good **representations**
- the ability to think in terms of **evaluation**.

Their materials are intended for all levels of school, but they also provide primary school specific resources and guides on CT for teachers, through *Barefoot Computing* [45]. Barefoot Computing gives an operational definition of CT. This covers the same set of CT concepts given in the above CAS description, and also a set of approaches that a Computational Thinker should use. These concepts and approaches are shown in figure 3.1. They provide in-depth descriptions of the concepts and approaches through their website.⁴

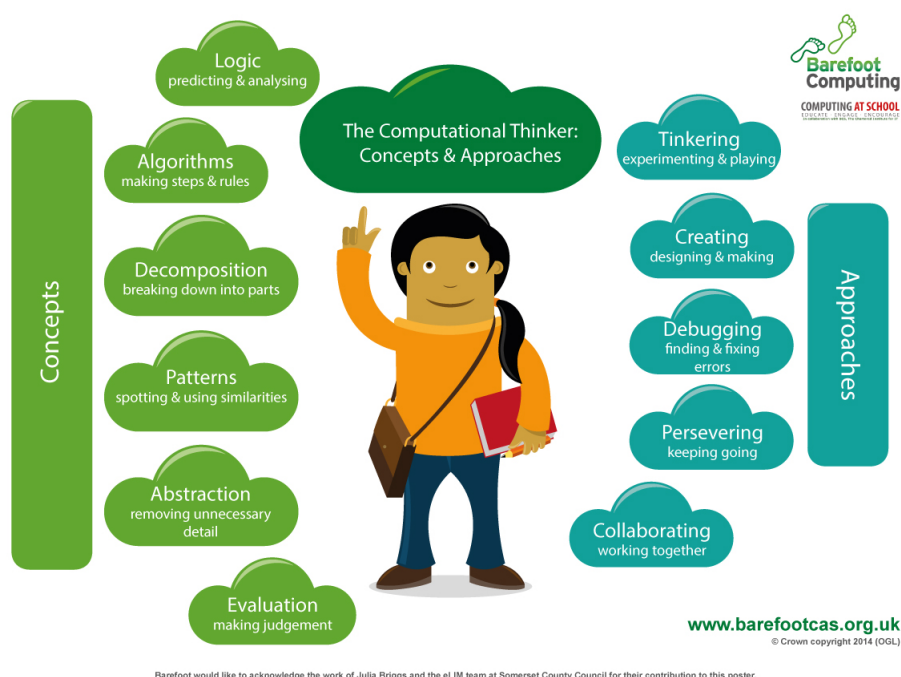


Figure 3.1: Computational Thinking Poster from CAS

⁴ <https://barefootcas.org.uk>

Chapter IV

Curriculum Developments

Over the past 10 years, Computer Science, programming, and Computational Thinking have been introduced to primary and high school curricula around the world. In many countries where this has not happened, there are plans to introduce these topics in the near future. In New Zealand, it was added to the high school Technology area of the curriculum in 2011, under the subject name Digital Technologies. It has since been introduced to primary school, and is expected to be fully integrated into all NZ schools by 2020.

In this chapter, I first give an overview of the many curriculum changes happening internationally, to illustrate just how significant a shift this is in school education. An outline of the education system and curriculum in NZ is then given in section 4.2, to situate the changes that have occurred, and how Digital Technologies and CS will fit into the National Curriculum. The events which led up to, and impacted the changes in the National Curriculum are documented in section 4.3. A curriculum framework, which was based on my work and proposed as a basis for updating the NZ curriculum, is discussed. Lastly, this framework's influence on the NZ Curriculum, and the current state of this curriculum is reviewed.

4.1 *International Curriculum changes*

Internationally, digital literacy and e-learning is generally seen as necessary in school education. Learning CS, on the other hand, has not been common in pre-tertiary education. There are a small number of countries, such as Israel and Lithuania [55], that have included CS and/or programming in their curricula for several decades already. However, for the majority of countries, including these topics in school curricula is a relatively new idea. Along with NZ the following countries have, or are currently, reworking their school curricula¹:

¹ This list is expanding rapidly, and new curricula will almost certainly have been announced since the publication of this thesis.

- **Australia:** CS, CT, and programming, were introduced to the Australian National primary curriculum in 2014, under the subject name ‘Digital Technologies’ [76]. This curriculum specifies what students should be learning from foundation (preschool), through to ages 15-16. It covers the areas data representation, algorithms, general computer use and digital literacy, digital systems, human-computer interaction, and programming.
- **Canada:** CS remains an elective subject in Canadian High Schools. Since 2010, CS has been a subject that can count towards University entry requirements, which has encouraged more schools to offer it [96]. Several provinces have made commitments to include and prioritise coding in their schools’ curriculum [77].
- **The EU (excluding the UK):** Throughout the EU, programming, CS, and CT, are part of school curricula in Austria, Bulgaria, the Czech Republic, Denmark, Estonia, Finland, France, Hungary, Ireland, Lithuania, Malta, Norway, Poland, Portugal, Slovakia, Spain, and Sweden [9]. Several of these countries have included this in the curriculum since the early 2000’s, while others, such as France, Spain, and Finland, did this more recently (2014-2016). It is also being considered for integration in Cyprus and Italy.
- **India:** There is currently no core curriculum for CS in schools across India, and ICT literacy has been the core computing related focus. However, there are initiatives focused on changing this, and while only a small number of secondary schools currently teach CS, this number is growing. One of the main challenges schools face in introducing CS, is a lack of networking infrastructure, and access to adequate computer hardware resources [145].
- **Japan:** The Japanese Ministry of Education is planning to introduce Computer Science to the primary school curriculum in 2020, with middle and high school curriculum to follow in 2021 and 2022 respectively. CS will be taught within existing subjects, rather than as a stand-alone subject. Before the introduction of this subject, the Ministry intends to provide both adequate ICT resources across all schools and regions. They also intend to provide professional development and lesson resources for teachers [183, 195].

- **South Korea:** Computer education has been present in the South Korean curriculum since the 70's. Since being renamed 'Informatics' in 2007, it has focused on CS concepts and principles. It had previously, however, been an elective subject, and participation in it had been low. In 2018, a new Informatics curriculum was released, made compulsory for middle schools, and made an elective in high schools. It covers digital literacy, CT, and programming [94].
- **The UK:** CS was commonly taught in UK schools during the 1980's, but over the next 20 years it was replaced with "Information and Communication Technology" (ICT). This ICT curriculum was focused solely on *using* digital technology. In 2012, The Royal Society released a report on the state of ICT in UK schools, which recommended large scale changes to the subject [79]. Since this was published, England introduced new computing standards to all school levels in 2014, followed by Scotland in 2016 [65]. However, there have been struggles with the implementation of the subject, as both England and Scotland have been unable to fill demands for computing teachers, and large numbers of these teachers are not confident teaching this curriculum. In Wales, a new curriculum is due to be released in late 2018/2019, which will include CS [176].
- **USA:** In January 2016 the then President of the United States of America (USA), Barack Obama, announced the "Computer Science for All" initiative, which aimed to provide all students in the USA the chance to learn CS in school [177]. Across the USA, many states and counties have added CS to their curriculum. CS standards exist in Ohio, Virginia, Utah, Wisconsin, Idaho, Indiana, Massachusetts, and Texas. Many more states have drafted standards, and pledged to introduce them in the near future [75, 117]. As well as new curriculum standards, the Code.org "Hour of Code" initiative has made a substantial impact in the USA. It claims that 30% of students in the USA have a Code.org account, and have completed at least one hour of coding through their website [40].

4.2 The New Zealand National Curriculum

The addition of Digital Technologies (DT), CS, and programming, to the New Zealand Curriculum, began in 2011. At the time of writing (2019), this has evolved to include CT, and extends through all levels of school. Before discussing these changes, it is useful to have an understanding of the NZ education system. In this section, the operation and structure

of the National curriculum, from the beginning of primary school to the end of high school, is explained.

4.2.1 *The New Zealand Curriculum, and Te Marautanga o Aotearoa*

In New Zealand there are two curricula: *The New Zealand Curriculum*², and *Te Marautanga o Aotearoa*³. *Te Marautanga* is a curriculum from a Māori perspective. These are two separate, and different curricula, but the majority of required subjects in each are the same. They are not translations of the same curriculum

My work took place in the context of the New Zealand Curriculum only, but for the subjects of CS, programming, and CT, there is a large degree of overlap between each curriculum. When referring to both of these curricula, the term *National Curriculum* is used.

These documents do not dictate the exact curriculum schools must follow. Instead they provide a framework around which schools build their own individual programmes. The New Zealand Curriculum serves as the framework for English-medium schools, and *Te Marautanga o Aotearoa* as a framework for Māori-medium schools (where *te reo Māori* is the language used for instruction - these schools are also called *kura kaupapa*). Schools must base their programs on the principles, values, and key competencies of their chosen curriculum, and through years 1-10 they must offer the subjects: English or Māori (depending on the school medium), the arts, health and physical education, mathematics and statistics, science, the social sciences, and technology. In years 11 to 13, students select the subjects they will continue with, and complete assessments that count towards future qualifications. The range of subjects available to students tends to expand in years 11 to 13, as students begin to specialise. Aside from these requirements, schools have a large degree of flexibility and freedom in setting their individual learning programmes, a relatively unique quality among school curricula.

The foundation of the NZ curriculum is a vision to produce “Young people who will be confident, connected, actively involved, lifelong learners”. The structure of the curriculum built on this is shown in Figure 4.1. ‘Learning Areas’ reflects the main subject areas students study. The vision of *Te Marautanga o Aotearoa* is similar. It “aims to develop successful learners, who will grow as competent and confident learners, effective communicators in

² <http://nzcurriculum.tki.org.nz/The-New-Zealand-Curriculum>

³ <http://tmoa.tki.org.nz/Te-Marautanga-o-Aotearoa>

the Māori world, healthy of mind, body and soul and secure in their identity, and sense of belonging. They will have the skills and knowledge to participate in and contribute to Māori society and the wider world”.⁴

4.2.2 The Key Competencies

A core part of the NZ curriculum is the five Key Competencies. Alongside subject knowledge, students are expected to develop these competencies, and use them throughout their learning. The NZ curriculum states: “The development of the competencies is both an end in itself (a goal) and the means by which other ends are achieved. Successful learners make use of the competencies in combination with all the other resources available to them. These include personal goals, other people, community knowledge and values, cultural tools (language, symbols, and texts), and the knowledge and skills found in different learning areas” [133].

The Key Competencies, and their official descriptions, are:

1. Thinking:

“Thinking is about using creative, critical, and metacognitive processes to make sense of information, experiences, and ideas.”

2. Relating to others:

“Relating to others is about interacting effectively with a diverse range of people in a variety of contexts. This competency includes the ability to listen actively, recognise different points of view, negotiate, and share ideas.”

3. Using language, symbols, and texts:

“Using language, symbols, and texts is about working with and making meaning of the codes in which knowledge is expressed.”

4. Managing Self:

“This competency is associated with self-motivation, a “can-do” attitude, and with students seeing themselves as capable learners.”

⁴ <https://parents.education.govt.nz/primary-school/learning-at-school/new-zealand-curriculum>

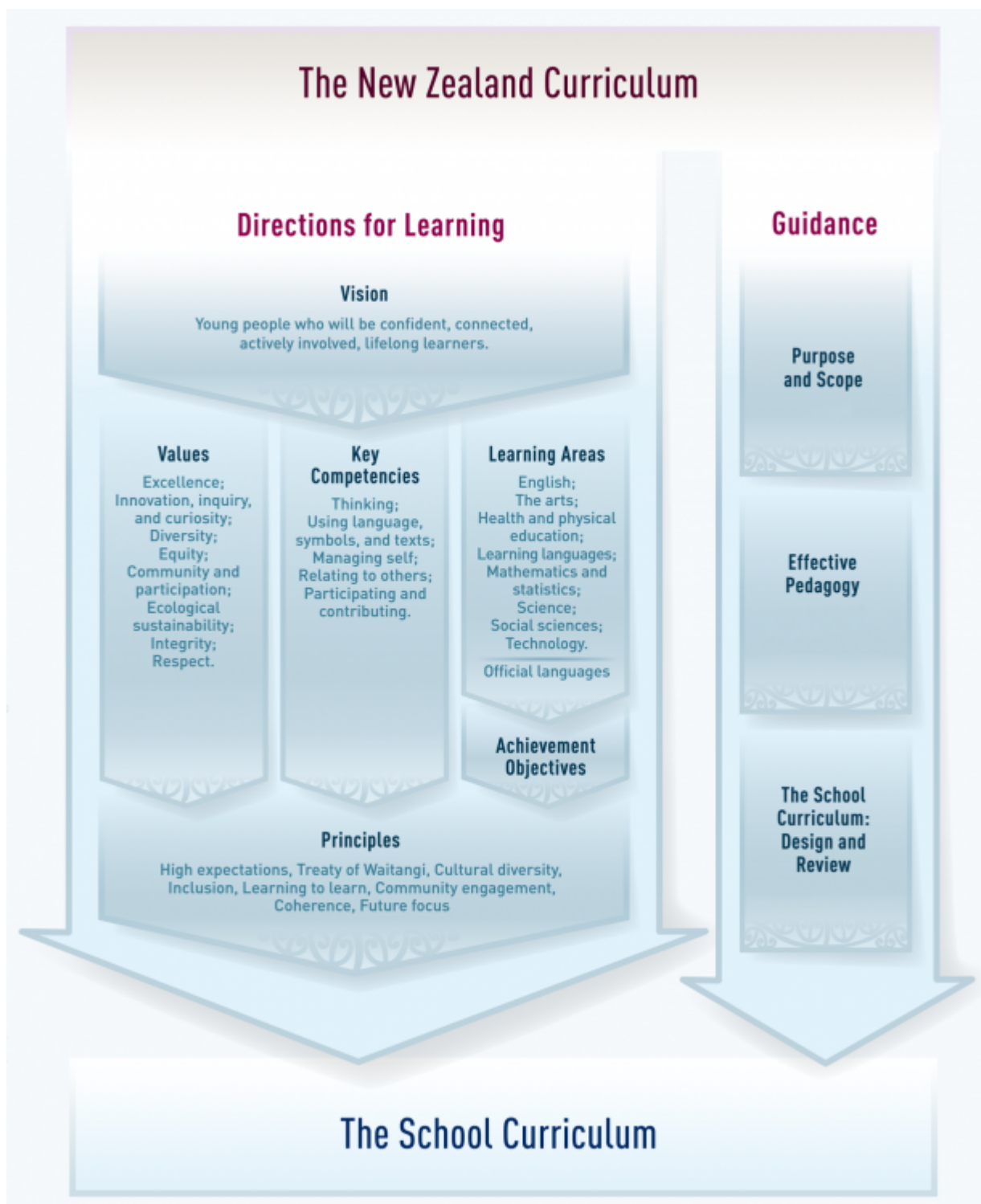


Figure 4.1: A Schematic View of the New Zealand Curriculum (New Zealand Ministry of Education)

5. Participating and Contributing:

“This competency is about being actively involved in communities. Communities include family, whānau, and school and those based, for example, on a common interest or culture.”

The relationship between these competencies and the topics of CT, CS, and programming will be revisited frequently throughout this thesis.

4.2.3 NCEA

The National Certificate of Educational Achievement (NCEA) is the most common high school qualification that schools offer in NZ. NCEA has 3 levels of assessment, with year 11 corresponding to NCEA level 1, year 12 to level 2, and year 13 to level 3. Students’ knowledge and work are assessed through Unit and Achievement standards, and they are awarded an amount of NCEA credits for each standard they successfully complete. These credits are required for entrance to tertiary education (including university and vocational institutions), and satisfying the pre-requisites for a number of tertiary courses. Unit and achievement standards differ, in that unit standards are pass/fail assessments, while students may obtain a range of grades in achievement standards. Students who pass receive one of the following grades, listed from lowest to highest: Achieved, Merit, or Excellence. Earning high grades in achievement standards can earn students scholarships, preferred entry to courses, and improved employment opportunities.

4.3 *History of Digital Technologies and Computational Thinking in the New Zealand Curriculum*

In this section the evolution of Digital Technologies in the National curriculum, and the events that triggered the changes to this at each step, are covered.

4.3.1 Pre-2011

Computers first became common in New Zealand schools during the 1980s. Savidan (2003) reviewed the developing situation of ICT in NZ schools through from the 1980s, to 2003 and concluded that, despite students generally positive attitudes towards this technology and the inclusion of ‘Technology’ in the 1993 New Zealand Curriculum Framework (NZCF), ICT had still not become an integral part of the New Zealand education system [155]. By 2003

secondary school students in the majority of NZ schools were using ICT, in some form, for learning. However, in the absence of guidelines for using it in curriculum, it was generally seen as a tool that teachers had to slot into existing curriculum. Students learnt *with* ICT, but generally not *about* ICT.

By 2008, this situation had not changed. In the vast majority of NZ schools, the subject of computing was restricted to using computer applications and file system management. The teaching of programming, CS concepts, and software design, was almost non-existent and generally only appeared in schools where a particularly interested and motivated teacher pushed the subject.

In 2008 two reports were published that denounced the state of computing in the National Curriculum, and urged the Ministry of Education and Government to take action [18, 37, 83]. Both reports highlighted the lack of suitable achievement standards that focused on computing, as a major issue for the subject. ICT unit standards were available, but these were pass/fail assessments. Unit standards are generally viewed by students (and frequently by teachers) as unsuitable for academically focused students. An alternative to using these unit standards was to apply some of the generic Technology achievement standards to ICT. But this approach had its own issues, as these standards were originally written with the intent of being applied only to physical materials and processes. The report from the New Zealand Computer Society (now known as IT Professionals New Zealand⁵) took all the existing Technology achievement standards that could be used for computing, and after in-depth critical analysis of these 18 standards wrote “we believe that **none** of the standards reviewed in this report are appropriate to assess either Computer Science or end-user computing at the secondary school level” (Grimsey & Phillipps, 2008, p.6). The second report, commissioned by the Ministry of Education, examined this from a broader view, discussing the negative implications maintaining the current curriculum could have for students and the NZ ICT industry (Carrell, Gough-Jones, & Fahy, 2008). They touched on the lack of gender diversity in the ICT industry, the need for all students to understand the digital world they are living in, and the ubiquity of computing in modern society and education.

The Ministry of Education responded to these reports by forming the Digital Technologies Experts Panel (DTEP), made up of representatives from the ICT industry, tertiary institutes and high schools. The panel was tasked with creating a plan to improve the NZ curriculum, by addressing the concerns that had been raised [18]. In May 2009 they recommended the

⁵ <https://itp.nz/>

creation of a new subject area, ‘Digital Technologies’, which would cover five areas, including programming and Computer Science.

4.3.2 The new standards, 2011 - 2013

In 2011 New Zealand became an early adopter of CS education in school curriculum, as the NCEA level 1 (corresponds to year 11 of school) Digital Technologies Achievement standards were first released and made available to schools. Achievement standards for levels 2 and 3 followed in 2012 and 2013 [180].⁶ At each level these cover core CS concepts and programming. These standards include the additional areas: digital information management and databases, digital media, electronics and embedded systems, digital infrastructures and networks. These areas are not covered in this research. All information covered in this section pertains to the versions of these standards released before 2014. The standards were reviewed in 2014, and at the end of 2018 these standards were replaced with ones reflecting the further changes happening in the NZ curriculum.

At level 1 and 2 there are two Achievement standards on programming and one at level 3, with each successive level requiring students to use more sophisticated programming constructs (e.g. functions, increasingly complex data types). There is one CS Achievement standard at each level. A student who completes all three of these standards will have covered the topics of algorithms, types of programming languages, human computer interaction, data representation, encoding, encrypting and compressing data. Like the majority of NCEA standards, these new Achievement standards were not compulsory for schools to offer. However, a significant number of schools have adopted these. In 2011, the first year these standards were available, 2701 students were enrolled in the level 1 programming standards, and 1429 in the level 1 CS standard [19]. As of 2016, this had grown to 9721 students in the levels 1 to 3 programming standards, and 3932 in levels 1 to 3 CS.⁷

4.3.3 Reviewing the Digital Technologies curriculum, 2014 - 2017

In 2014 the New Zealand Government released the report “A Nation of Curious Minds - He Whenua Hihiri i te Mahara. A National Strategy Plan for Science in Society” [127,

⁶ These standards, and the updated versions which have since been released, can be accessed through the NZQA website <http://www.nzqa.govt.nz/ncea/assessment/search.do?query=Digital+Technologies&view=all&level=01>.

⁷ <http://nceatechstats.com/digital/>

173]. This reviewed the current situation of science and technology education in NZ, with a large part of this focused on computing. The review recommended an evaluation of the Digital Technology curriculum, and consideration of the place of computing in the overall curriculum. This opened the door for the development of a DT curriculum that would extend into primary school. Throughout 2014 and 2015 a review of “The Positioning and Content of Digital Technology within the New Zealand Curriculum/Te Marautanga o Aotearoa” took place, with frequent working groups being held with key stakeholders in this area. These stakeholders included representatives from the New Zealand Association of Computing and Digital Technologies Teachers (NZACDITT, now known as Digital Technologies Teachers Aotearoa, DTTA), the NZ Institute of IT Professionals, the technology Industry, educators (from both English and Māori medium), the Ministry of Education, NZ Universities, and many associations and companies focused on Technologies education in NZ (both English and Māori medium). Professor Tim Bell and I were members of this review team. In October 2015 Professor Bell and I submitted a proposal to the review team for a primary school curriculum to be created, and the subjects for it to cover at each school level. This was based on our work researching international curricula and on my work with NZ primary teachers [69]. Much of the content in this proposal subsequently became part of the new Digital Technologies curriculum, launched in 2018. It also matched the majority of the curriculum content I used throughout my work with NZ teachers.

As a result of this review Hekia Parata, the New Zealand Minister of Education at the time, announced on July 5, 2016, that “Digital technology [was] to be formally integrated into the New Zealand Curriculum and Te Marautanga o Aotearoa” [174]. Parata stated the change would “ensure that we have an education system that prepares children and young people for a future where digital fluency will be critical for success”. This meant CS, and programming, which had previously been restricted to an optional subject in the final three years of school, would be extended to cover all year levels. Computational Thinking would also be incorporated, explicitly, into these subjects. Curriculum development was to take place from this point, until the end of 2017, with a goal of having the subject integrated into the National Curriculum in 2018.

On June 28, 2017, Nikki Kaye, the then New Zealand Minister of Education, announced the launch of the consultation on the new digital technologies content for the New Zealand Curriculum and Te Marautanga o Aotearoa [175]. In her announcement speech she stated “We’re breaking new ground with a curriculum that offers unique Māori content, learning that can be shaped according to students’ individual needs, and future-proofing so it can

adapt to new technology as it arises... The new curriculum content is about ensuring that students across all year levels have access to rich learning aimed at building their digital skills and fluency, to prepare them for this world... All young people from years one to 10 will take part in digital technologies learning. Students choosing digital technologies pathways for NCEA will develop the more specialised skills that industry partners say are in high demand, through new achievement standards being developed for NCEA Levels 1, 2 and 3”.

A month later the draft curriculum⁸ was released for consultation.⁹ As previously planned, the new curriculum content was made available for the 2018 school year, but with a planned two year transition period. The curriculum is planned to be in full use in all schools from the start of 2020.

4.4 Curriculum Content for New Zealand, and this research

As previously mentioned in section 4.3.3, Professor Tim Bell and I were part of the 2014 to 2015 NZ Digital Technologies curriculum review team. During 2016 and 2017, I continued to participate in Ministry workshops, discussions, and the writing of the new curriculum. Much of the work discussed in this section was conducted as part of Professor Bell’s and my contribution to this review, and the work reported on in our 2015 paper “A Pilot Computer Science and Programming Course for Primary School Students” [67]. This work has been a strong influence on the new NZ primary school curriculum.

4.4.1 Synthesising a curriculum

When the writing of the new curriculum began, several countries had already developed and released CT curricula at primary and post-primary school levels. In 2015 I examined a range of these curricula, covering primary through to tertiary education, and CT curriculum reviews, to identify common themes and synthesise a proposed NZ curriculum. These were:

- England’s computing curriculum, covering key stages 1 to 4 (from 5 to 16 year olds) [65],
- Australia’s Digital Technologies curriculum [7, 76],
- the CSTA K-12 Computer Science standards [157],

⁸<https://education.govt.nz/assets/Documents/Ministry/consultations/DT-consultation/Digital-Technologies-Hangarau-Matihiko-ENG-July.pdf>

⁹<https://www.curiousminds.nz/news/have-your-say-on-digital-tech-education/>

- the NZ Digital Technologies NCEA Achievement Standards [134],
- the AP Computer Science Principles programme [109], and
- additional reports and publications on CT curriculum [15, 122, 170], computational participation [105], and computational learning [46], from experts in the field.

Several of these have since been updated.

The topics and themes identified in these documents had much in common, and often in cases where they disagreed there was a distinct difference in the meaning behind terms used in each document (e.g. differing views on what “digital literacy” encompasses). The comparison of the Australian, English, and CSTA curricula can be found in Appendix A, tables A.1, A.2, and A.3. After comparing these documents we settled on the following topics from which to create a curriculum. These are taken from our paper in the Proceedings of the 2015 Workshop in Primary and Secondary Computing Education [67].

Algorithms: includes articulating a plan for a program; exploring standard problems such as searching or sorting; computational complexity (performance); design techniques (such as decomposition and abstraction).

Programming: includes implementing a plan/algorithm, debugging, and testing; selected programming environments e.g. “Initial Learning Environments” (drag and drop), general purpose languages, object oriented programming; parallel execution.

Data representation: includes binary representation; types of data — numbers, text, sound, images etc.; encoding the data — compression, error correction, encryption; data structures.

Digital devices and infrastructure: includes components, such as CPU, memory, data storage devices; input and output devices; networks; cloud computing; troubleshooting devices and configurations.

Digital applications: includes simulation and modelling the real-world; creating digital content (media) such as documents, images, presentations, web pages, video; publishing information; collecting and interpreting data (spreadsheets, databases); intelligent systems.

Humans and computers: includes digital citizenship (cybersafety, privacy, intellectual property including copyright and licensing, sustainability, ethics, equity, accessibility, legal responsibilities); careers (including diversity); project management; usability; connection to other fields; keyboard proficiency.

A curriculum framework, based on these six topics and the results of the pilot study I conducted in 2014 (covered in Chapter 7), was proposed to the NZ Digital Technologies review group. Two versions of this framework were presented, a *conservative* version and a more *ambitious* version. These each covered broadly the same content, but the *ambitious* version introduced content earlier in primary school, and added extra subtopics in several areas. These two frameworks can be found in Appendix A, tables A.4 through A.7. For the work I conducted from 2015 to 2017 I used these frameworks as a reference model for the resources, professional development, and guidance I provided teachers with.

4.4.2 Current and future NZ Curriculum

As discussed in section 4.3.3, the changes to the National Curriculum were announced in 2018. The Technology/Hangarau learning area has been modified in the NZ curriculum and in Te Marautanga o Aotearoa so that it is now split into five different *strands*¹⁰, two of which are specific to Digital Technologies/Hangarau Matihiko¹¹.

In the NZ curriculum these strands are:

- **Computational Thinking:**

Students will develop and understanding of computer science principles that underlie all digital technologies. They'll learn core programming concepts so that they can become creators of digital technology, not just users.

- **Designing and Developing Digital Outcomes:**

Learning how to design quality, fit-for-purpose digital solutions.

¹⁰ <http://nzcurriculum.tki.org.nz/The-New-Zealand-Curriculum/Technology/Learning-area-structure>

¹¹ <https://www.education.govt.nz/ministry-of-education/specific-initiatives/digital-technologies-and-hangarau-matihiko-learning/>

In Te Marautanga o Aotearoa they are:

- **Te Whakaaro Rorohiko** (*Computational Thinking*):

Includes using te reo Māori to express problems, formulate solutions and solve them using algorithms, programme and data representation.

- **Tangata me te Rorohiko** (*People and computers*):

Includes designing and developing digital outcomes while considering their role and responsibility as digital citizens.

The details of both curricula can be found through the Ministry of Educations website¹².

¹²<http://education.govt.nz/ministry-of-education/specific-initiatives/digital-technologies-and-hangarau-matihiko-learning/>

Chapter V

Teaching Computer Science and Programming in Primary School

In this chapter I discuss some of the most prominent resources that are available for teaching Computer Science and programming to primary-aged students, and the research related to these. This includes the Computer Science Unplugged programme, its use internationally, and how it can assist students development of Computer Science knowledge. I then discuss tools and programming environments that can be used, and those that have been developed specifically for primary aged children.

As the Computer Science and programming concepts to be taught during this research were identified in Chapter 4, and the training provided to teachers on how to teach these are covered in Chapter 6, they are not discussed here.

5.1 Computer Science Unplugged

A widely used and successful strategy for introducing CS to young students is *Computer Science Unplugged* (CS Unplugged).¹ Computer Science Unplugged is a collection of activities and lessons that can be used to teach CS concepts without using computers, or any other type of digital technology. The activities require no prior knowledge of CS and can be completed using common classroom supplies, such as paper, pencils, or chalk. CS Unplugged follows a constructivist approach to learning and promotes collaboration among students. It was originally created and designed to be used for ‘one-off’ sessions with primary aged children, with the goal of raising awareness and interest in CS, rather than as part of a curriculum. However, it is now widely used internationally for both of these purposes, at both pre-tertiary and tertiary level.

It has been discussed and recommended in CS education books [39, 93, 160], by the

¹ Classic CS Unplugged: <https://classic.csunplugged.org>, and the updated version of CS Unplugged, released in 2016: <https://csunplugged.org>

prominent CS and programming initiative Code.org,² and has been recommended for use in curricula by the Association for Computing Machinery (ACM) [182].

The activities were originally developed in the 1990s by Tim Bell, Mike Fellows, and Ian Witten. A book of these activities, “Computer Science Unplugged... off-line activities and games for all ages” was published in 1998 [22]. In 2000, 12 of the activities were adapted for classroom use by teachers Robyn Adams and Jane McKenzie [20], alongside the original authors. From 2000 to 2012, the remaining activities were adapted for classroom use and released [21].³ In 2016, the UC Computer Science Education Research Group began redeveloping the activities, with the aim of generating a larger series of lessons, based on each activity, for use as in primary school curriculum.

The CS Unplugged materials are available under a Creative Commons Attribution-ShareAlike licence. This allows others to use, modify, and build on these materials (with acknowledgement). Because of this, there are many variations of the original CS Unplugged activities. Many different activities and lessons that follow the ‘unplugged’ approach of teaching CS without computers have been created by a range of organisations and people. The term ‘CS Unplugged’ is now often used to describe any activities that are intended to teach CS or programming concepts without using computers. Unless otherwise stated, I will be referring to only the original CS Unplugged activities (now known as ‘Classic’ CS Unplugged) as they are described in [21] and [22], and the updated version of these original CS Unplugged activities, first released online in 2016 [172]. The activities and additional content provided through the new CS Unplugged website have been continually updated since the website went live, as classic CS Unplugged activities have been adapted for curriculum use and added to this new version.

Research on the use of CS Unplugged varies between using the original set of activities, variations on these, new activities not in the original set, or combinations of these. The original activities and similar variations on these are frequently found to be highly engaging, appeal to a wide range of people and age levels, and increase students’ interest and confidence in CS [67, 68, 69, 113, 148, 178, 179]. It has been shown to be more or equally as effective in teaching the intended CS concepts when compared to traditional and alternative methods [113, 147, 179], and works effectively when combined with teaching programming. It has also been used and adapted to teach students with disabilities, in an inclusive manner [47].

² <https://code.org/>

³ The full history of releases can be found at <https://classic.csunplugged.org/changelog/>

There have been critiques of the classic CS Unplugged activities, which suggest that they are not sufficient for use as part of a school curriculum. For example, Theis and Vahrenhold [178] investigated whether these outreach activities were suitable for use as teaching resources in primary school CS courses. They took 12 of the original activities, as they appear in the first teachers edition [20], extracted learning outcomes from these, and mapped these to Blooms revised taxonomy [4]. Theis and Vahrenhold identified that the core levels on this taxonomy which CS Unplugged did not cover were ‘Evaluating’ and ‘Creating’. Both of these are included as key components of Computational Thinking in a large number of the different CT definitions, including the Barefoot Computing framework I used in my studies. Theis and Vahrenhold stated extending the activities, so they include elements of evaluation and creation would improve their use for CT education. They also found that the activities did not extend to ‘meta-cognitive knowledge’, but note that this is not problematic, as children are not expected to be using this knowledge at primary school ages [141]. In their work, the authors acknowledged that, as outreach activities, the activities they analysed were not written explicitly to include learning outcomes, or to have any outcomes extend across all levels of the taxonomy.

Rodriguez investigated whether CS Unplugged activities could successfully develop CT skills in middle school (ages 11-13 approximately) students [148]. The activities used in their work had been extended from the classic versions based on teacher and student feedback, and a final project for students to complete was written. After these modifications, the activities extended to higher levels of Bloom’s taxonomy and had more advanced learning objectives which were suitable for middle school students. Based on students results in the final project, Rodriguez concluded there was evidence that the extended CS Unplugged activities developed students CT skills [148].

Building on this work, Rodriguez, Kennicutt, Rader, and Camp reported on the results of administering a CS Unplugged based curriculum with 7th grade (ages 12-13) students and assessing their CT skills with pre and post-tests [147]. This curriculum included the activities described by Rodriguez [148], with some improvements. Their findings showed that students were effectively learning about the Computer Science concepts intended to be covered by these CS Unplugged activities. Their findings also lend support to the argument that CS Unplugged can help students develop CT skills. However, as they state in their work, these findings are not fully conclusive: “CT skills are fairly broad, so it is clearly not possible to claim that students have mastered a particular CT component based on just one assessment” [147].

The issues raised in these critiques provide insights into how the original activities could be modified and made suitable for primary school curricula. In particular, they highlight the need to extend the original activities so they explore topics more deeply, and to include learning outcomes. Extensions that could be taught in a variety of contexts would help students transfer the concepts they learn to different applications. They also emphasise the need for activities that include evaluation and creation and generally cover a wider range of cognitive processes. These issues were key drivers of the redevelopment of CS Unplugged, and the release of new resources and lesson plans, available at [172]. This new version of CS Unplugged has not yet been shown by others to succeed in addressing these needs. However, the results of my work, discussed in Chapters 8 and 9, provide evidence that the lessons from this new version of CS Unplugged do cover these areas.

5.2 *Programming for young students*

If programming is to be taught in primary schools, to diverse groups of students, it is important that it is taught in an inclusive fashion. Historically the way programming has been taught has often not been inclusive; it was generally more appealing and tailored to the interests of particular groups of students [123]. Much prior work reports on small groups of students becoming extremely interested and focused on these subjects and learning well, but does not focus on wider groups of students, or report on their general interest and success. Additionally, research on programming education has, until recently, very rarely focused on typical school environments. The majority of work has focused either on adults (in tertiary and non-tertiary environments), or self-selected groups of school age students, for example at after-school clubs. These environments are very different from that found in a primary school classroom.

Some of the earliest work on teaching young students programming, which did not take place with a self-selected group of students, was conducted by Seymour Papert. In *Mindstorms* [139], Papert wrote about the programming language Logo and using it to teach geometry based programming. Beginning in the 1960s, he worked with students from a range of age groups and ability levels, and concluded that “children of almost any age could learn to program in Logo under good conditions with plenty of time and powerful research computers”. The requirements of ‘good conditions’, ‘plenty of time’, and powerful computers, however, emphasises how programming education was restricted to a small minority of children at the time. Those who had access to computers, a knowledgeable programming

teacher, and already had a strong interest in computing to devote time to it, would have been some of the only students to have the opportunity to learn in ‘good conditions’. Alan Kay also reported on the success he had observed middle school students have when learning programming using the educational programming language SmallTalk in the 1970s [106]. However, as Kay wrote in 1993 when reflecting on this work, the success he had observed was likely overstated [106].

“The successes were real, but they weren’t as general as we thought. They wouldn’t extend into the future as strongly as we hoped. The children were chosen from the Palo Alto schools (hardly an average background) and we tended to be much more excited about the successes than the difficulties. In part, what we were seeing was the ‘hacker phenomenon’, that, for any given pursuit, a particular 5% of the population will jump into it naturally, while the 80% or so who can learn it in time do not find it at all natural.”

- Alan Kay, 1993, *The Early History of SmallTalk*

When the first portable computers started entering the home in the late 1970s and early 80s, the opportunity to learn programming did expand to more children, particularly as these early computers required the user to program them at times. However, this opportunity was yet again restricted to a small, financially privileged, group of society.

Even today, with computers accessible to students in schools, these ideal conditions are unlikely to exist in a typical primary classroom. In NZ, primary teachers are already hugely time constrained without having to add in another subject to the ‘crowded curriculum’. In NZ there are initiatives working to train teachers and provide high quality teaching resources, but the majority of primary school teachers are still not trained in delivering programming lessons. Now that there is a demand for widespread programming education for children, we must establish how programming can be taught in an inclusive way within a school environment.

One of the key goals of introducing these new subjects is engaging a diverse range of students in them, and increasing their interest in pursuing careers in this, or related, fields. This is particularly important for primary school education, as being introduced to CS and programming after primary school age, rather than earlier, makes it less likely that students will gain confidence or an interest in this subject. In teaching CS and programming, it is important to consider both how we can engage these students, but also how we avoid potentially discouraging them from pursuing these subjects. This is another issue that

appears to be particularly important for girls, as early negative experiences in computing have a particularly powerful impact on them [123]. It would, of course, be both unrealistic and undesirable to have *all* students decide that computing was the career path for them, and some students are unlikely to ever consider this a subject they want to continue with. However, when students decide that they do not want to pursue the subject, we want this to be for the right reasons, e.g. it's just not a field they are interested in or enjoy. We do not want students to be put off by, for example, influences like bad learning experiences, stereotypes and misconceptions about the field and the 'type of people' who can, and can't, succeed in computing.

The age at which students begin learning programming has been a topic of debate in media, curriculum development, and CS education research. This topic was explored in a paper I co-authored in 2014 [69]. Many of the arguments against programming being taught to young students revolve around it being an unnecessary skill for them to learn. It is seen by some to be a skill that will not be a required skill in the future, and focussing at such an early age on a skill that is specific only to some career paths is a waste of class time. The argument against this is that the goal of including this in curricula is not simply to produce programmers, but by learning it students will develop CT skills, problem solving skills, and a deeper understanding of the digital world.

5.2.1 Programming tools for young students

One of the most important considerations in teaching introductory programming is what tools should be used, and which tools are better suited to different age groups. Through the studies I conducted, it was clear from the teachers feedback that tangible learning materials were necessary for the youngest learners (approximately years 1-4, ages 5-9), and also highly beneficial for older students. Similar results for very young students were found by Kazakoff and Bers, who reported on the success kindergarten aged children (3-6 years old) had with using robotic toys to learn about sequencing [107]. Most of the youngest students (approximately years 1-2, ages 5-7) in my study, and all older students were able to use visual programming languages (also referred to as block-based, and graphical programming languages), such as Scratch. Some of the more advanced students in years 7-8 (ages 11-13) progressed to text based languages such as Python and JavaScript, but these students were generally those who had prior programming experience, and then worked on this independently. As the development of abstract thinking and spatial reasoning skills most commonly

happens throughout ages 11-16 [142], this need for tangible learning materials and visual, as opposed to text based, programming languages makes sense. It is worth noting that, to my knowledge, there were no participants in my studies with vision impairments, and so this was not a factor in the choice of programming languages students used.

Conceptualising the way a program functions without some type of tangible or visual feedback is difficult for younger students. In 2013 Salleh et al. completed a review of papers published 2005 and 2011 on research into programming tools for teaching, with a particular focus on introductory programming [154]. They discussed the strong impact a programming environment (meaning both the language and development environment used) can have on students ability to learn, and their continued engagement with learning. They emphasised that “In teaching and learning of programming, programming tools is one of the main topics that discuss issues related to pedagogy, curriculum and programming languages” [154]. They found that the tools used in tertiary education and in industry are not suitable for teaching primary aged students, and tools that are designed specifically to facilitate learning should be used instead. Primary aged students also require much simpler developer environments than those used by adult programmers and learners, as more advanced environments add an extra level of complexity to learning, and their extra capabilities are superfluous in primary school.

Many programming and ‘coding’ resources directed at primary aged children have been released over the past 5-10 years. These have provided a plethora of programming tools that are suitable for primary school education. The creation of these is largely due to the influence of not only the growing number of countries implementing curriculum, but also from industry interest. Many companies in the computing technologies sector have become engaged with CS and programming education, largely motivated by the current shortage of trained professionals in this area, and their desire to attract more people to the field. These resources include ‘Learn-to-Code’ websites, apps, and online programmes (e.g. Code.org⁴, Code Club Aotearoa⁵, Codecademy,⁶ Kodable,⁷ and Tynker⁸); programming languages de-

⁴ <https://www.code.org>

⁵ <https://codeclub.nz/>

⁶ <https://www.codecademy.com>

⁷ <https://www.kodable.com/>

⁸ <https://www.tynker.com/>

signed specifically for beginners and young learners (e.g. Scratch⁹ and Blockly¹⁰); single-board computers and microcontrollers (e.g. Raspberry Pi,¹¹ and Arduino boards¹²); and robotic toys (e.g. Bee-Bots¹³ and Spheros.¹⁴) Resources of these kinds were used by the majority of participants in my studies, to varying degrees.

When the Learn-to-code movement first started to take off, a popular way to introduce students and teachers to coding was the ‘Hour of Code’, an annual event run by Code.org which began in 2013. Having students complete one hour of coding through tutorials on Code.org became extremely popular, particularly in the USA. According to Code.org, 30% of students in the USA have now participated in this initiative. Other initiatives based on this model of short introductions to coding, were also launched. While this has greatly raised awareness of the push to introduce coding to schools, it was extremely limited in terms of actually developing students’ programming skills, and created issues for teachers. After completing these types of activities, students were frequently energised and wanted to continue learning. However, through the discussions I had with primary (participants and non-participants in my studies) and high school teachers during my work, it was clear that the majority of teachers were unsure what the next step in teaching was after their students had completed the hour of code.

The focus on ‘code’ rather than programming and CS was also seen by some to be problematic. The words coding and programming are frequently used interchangeably, but this can create confusion [69]. In the field of CS and computing, programming generally refers to the full process of taking a problem or specifications, coming up with an algorithmic solution to this, and implementing it in a programming language. Coding has a large number of meanings in CS that are entirely unrelated to programming, but when it is used in the context of programming it is typically used to refer only to the final step, of taking the already created algorithm, or pseudo-code, and transferring this to a programming language [69]. A large number of these introductory activities, and also longer Learn to Code online programmes, follow this model of simply translating a pre-written algorithm into code. This is

⁹ <https://scratch.mit.edu/>

¹⁰ <https://developers.google.com/blockly/>

¹¹ <https://www.raspberrypi.org/>

¹² <https://www.arduino.cc/>

¹³ <https://www.bee-bot.us/>

¹⁴ <https://www.sphero.com/>

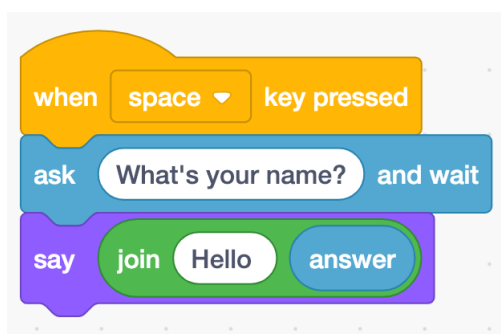


Figure 5.1: A program written in the visual programming language Scratch (version 3)

not to say these types of lessons are not beneficial - they have been compared to learning to read before learning to write; but alone they are not enough to fully teach the process of programming and CT skills.

In order to address these limitations, since 2013 Code.org has introduced many CS activities; a set of programming courses, called ‘Code Studio’, for different school year groups; and a large collection of resources for educators. The majority of popular Learn-to-Code websites and programmes have also done this. This has provided teachers with follow up material that they can continue to engage their students with. These programmes also take a large amount of pressure off teachers as they provide pre-prepared lessons and projects, and allow students to work at their own pace. They were not covered in the training that we conducted with teachers, but teachers were able to use them if they felt it would suit their teaching programme.

Visual programming languages are frequently used to introduce students to programming, and may be more effective for learning than text-based languages [48, 154]. These were used by the vast majority of teachers I worked with. A visual programming language is a programming language that allows the user to construct programs by choosing and connecting programming elements (such as loops, variable declarations, and conditional statements) that are displayed visually, as opposed to typing commands. A short Scratch program composed of these visual commands is shown in Figure 5.1. These commands are frequently referred to as ‘blocks’.

There are pros and cons to using visual programming languages as teaching tools. They can simplify the process of programming by providing a clear set of programming instructions for students to choose from and eliminating syntax errors [126] They are also highly engaging, and the immediate feedback they give is generally easier for children to interpret than that

given by a traditional programming language. Teachers with little programming experience find these languages less intimidating than text-based programming.

On the other hand, Meerbaum-Salant, Armoni, and Ben-Ari (2011), observed that students using Scratch frequently developed undesirable programming habits and used an intensely bottom-up programming approach [126]. An analysis of programs in the online Scratch repository found that the vast majority of projects that had been created demonstrated little to no understanding of programming concepts by the creators, and showed the use of these negative programming habits as well [2]. Meerbaum-Salant, Armoni, and Ben-Ari noted that the development of these habits happened when students used Scratch in an open-ended and exploratory way [126]. They observed that when students were explicitly taught concepts they did create projects which made use of these. Their work suggests that visual programming languages are a suitable tool for teaching students programming, but that they are specifically a *tool*, and explicit teaching of programming concepts is also required.

There is little research on students ability to transition from visual programming languages to text-based ‘real’ languages. Results of this research also seem to disagree. Some indicates that this is a difficult transition for students, and learning with visual languages can even be detrimental when students move to some text-based languages [143]. Other work has found students who have previously worked with visual programming languages grasped programming concepts in text-based languages quickly and gained a more in depth understanding, when compared to students who had not [5]. These students also displayed more enthusiasm for programming. Despite these observed differences in students speed of understanding however, there was little difference in these two groups’ achievement when assessed. As I worked only with primary school aged students, and text-based languages are generally introduced at high school instead, the transition between these was not a topic I investigated in my work.

5.2.2 *Programming books and games*

Along with programming environments and online courses, programming books and physical (i.e. not computer based) games have also been developed. These provide an alternative way to engage students with foundational programming concepts.

There are textbooks and activity books available for teachers that give guidance and lesson plans for teaching specific programming languages, and example projects for students to

complete. For example, “How to Teach Primary Programming” by Phil Bagge [8], published by Code-IT Primary Programming, gives a study programme for teaching CS with Scratch. It was written to support the English Computing curriculum (specifically Key Stage 2, ages 7-11), but covers topics common to many curricula, including New Zealand’s. The book contains Scratch projects for students to complete, that are intended to cover the CS and programming topics in the English curriculum, and gives guidance on debugging and general approaches to teaching programming. More books on Scratch programming have also been published by Creative Computing.¹⁵ These are free, and include guides for educators and a workbook for students that covers both Scratch programming and the design process [27, 28]. These books have received high praise from educators who have used them. Books for self directed learning of text based languages are also available for primary aged children, such as Jason R. Brigg’s books on Python programming, “Snake Wrangling for Kids, Learning to Program with Python” [30], and “Python for Kids. A Playful Introduction to Programming” [31]. These are useful for keeping more advanced students, who may be looking for additional challenges, engaged in the learning process.

Storybooks have also been written for teaching programming. “Hello Ruby”, by Linda Liukas [120], is an illustrated children’s book in which “kids will be introduced through storytelling to the basic concepts of coding”. The book also contains many ‘coding’ activities which, like CS Unplugged, do not require computers and are play based. It covers CT concepts along with basic programming. Storytelling is also used to explore CS and computation in the book “Computational Fairy Tales”, by Jeremy Kubica [112], and the additional fairy tales he publishes online [111]. These fairy tales cover a large range of computing topics, including algorithms, data structures, programming, Boolean logic, and compilers [111]. These resources were not used in my studies, but several teachers discovered similar books themselves and used these with their students.

Another alternative to learning programming through online games and apps is using physical games, such as card games and board games. For example, the board game *Robot Turtles*,¹⁶ advertised as ‘the game for little programmers’, involves constructing programs from game cards to move your turtle from place to place on the board and collect gems. It was inspired by Logo, and teaches simple sequencing in a similar way to using Bee-Bots, and doing the CS Unplugged KidBots activity. Games covering more advanced programming

¹⁵<http://creativecomputing.gse.harvard.edu/guide/index.html>

¹⁶<http://www.robotturtles.com/>

concepts have also been released. These include *Potato Pirates*,¹⁷ in which players can create programs with loop, conditional, and function cards. Another example is the marble-based logic puzzle solving game *Turing Tumble*,¹⁸ that contains enough programming concepts to qualify as a Turing-complete computer. The player places different components, such as cogs and ramps, onto a board to create a system that will produce a specific pattern of coloured marbles. Placing components on the board allows you able to simulate conditionals, Boolean logic, loops, and memory. These games can also be combined with storytelling. *Turing Tumble* includes a comic book with puzzles to complete (advertised as challenging even for adult programmers), and an interactive *Robot Turtles* eBook is available. Physical games like these can be used to provide more hands-on and less abstract ways to learn programming.

¹⁷ <https://www.potatopirates.game/>

¹⁸ <https://www.turingtumble.com/>

Part II

Trialling in New Zealand Primary Schools

Chapter VI

Methodology

In this chapter, I discuss the three studies I conducted through 2014-2017, the context these took place in, and my research approaches. I took a pragmatist approach in my methodology, used a mixed methods approach to data collection for each of these studies and employed a combination of descriptive quantitative, quasi-experimental quantitative, and phenomenological qualitative research methods. The studies were conducted with NZ primary school teachers and educators, trialling Computer Science and programming lessons in their classrooms. I trained the participants in using Computer Science Unplugged and block-based visual programming languages to teach basic Computer Science and programming concepts.

In this chapter, I first provide an overview of the three studies I conducted through 2014-2017, and describe the typical classroom environment in NZ, the NZ school decile system, and priority learners. I then discuss the teaching resources and training that participants were provided with, and how these changed throughout my research. The methodology of my data collection and analysis is covered, and the limitations and threats to the validity of the research are discussed. Finally, details on the ethics approval obtained for these studies, and on discussions of gender are given.

6.1 Overview of studies

In 2014 I began my research by conducting an exploratory Pilot study, working with one teacher at an intermediate school, with approximately 600 students. During this time I worked closely with this teacher and supported her to integrate Computer Science and programming lessons into her classroom programme. I collected feedback on both hers and her students' experiences with the lessons; information on students' attitudes towards the topics; the students' self-perceived aptitude for the topics; and students' scores in three tests. This work was partially reported on in a paper I co-authored, in 2015 [67].

I then conducted a wider study with 40 teachers, throughout 2015 and 2016. During this time I gathered feedback on the lessons they taught. This work and the preliminary results of it were reported on in a paper I co-authored, in 2017 [68]. The majority of the participants were full-time primary school teachers. The other participants were relief teachers, and people who were not registered teachers but worked in the Digital Technologies education field. This group of 40 all participated to varying degrees, and at different times throughout the 2 years the study ran. During this study support and training were offered to participants through workshops conducted by the UC Computer Science Education Research Group (CSERG), and one-on-one tutoring and classroom support from myself and one of my colleagues, Wendy¹, in the CSERG. Wendy's involvement is discussed in more detail in Chapter 8.

In 2017 I ran a more targeted intervention study. I worked with 3 schools, and a group of teachers from each one. These teachers taught their classes specific lessons I assigned to them, while other classes in the school did not receive these lessons. I conducted a pre and post test on CT with all classes at each school, and interviewed the participating teachers.

Ethics approval was obtained for each of these studies from the University of Canterbury Educational Research Human Ethics Committee² before commencement. Details can be found in Appendix C.

6.2 Teaching in a New Zealand primary school environment

In NZ, a *typical* primary school class would: be taught by a generalist teacher, who has trained specifically in Education and Teaching [36]; contain between 17-30 students, with the most common size being 28 students [188]; have access to high-speed internet and computers (through school equipment, 'bring your own device' (BYOD) programmes, or a combination of the two) [99]; contain students with a large range of ability levels across different subjects; and follow either the NZ Curriculum or Te Marautanga o Aotearoa. In NZ, schools that have students with high-needs or special education needs will typically include these students in general age-appropriate classes, rather than having separate classes [74]. In 2015 90% of schools had at least one high-needs student enrolled [74]. It is therefore not unusual for a class to include one or more high-needs students. The description 'high-needs' covers students with, for example, severe behaviour problems, moderate to severe physical disabilities,

¹ Name has been changed

² <http://www.canterbury.ac.nz/study/ethics/educational-research-human-ethics-committee/>

communication difficulties, and cognitive development delays [132].

If CT and CS (which come under the title of Digital Technologies in NZ) are to be successfully integrated into primary schools, it is crucial that they are taught in a way that is accessible for all students and generalist teachers without prior CS experience [88]. In the past, this has often not been the case, and in many cases the teaching methods used for computing education have been found to be biased towards different groups of students [123], or only investigated in situations with students who were self-selected and already enthusiastic [69, 72, 163]. As the technology industry is struggling with a lack of diversity in its community, teaching practices need to make CS and computing accessible to students regardless of their gender, ethnicity, socio-economic background, disabilities, and so on.

For any subject taught in a primary school environment, there are additional considerations that need to be made. Teaching materials need to be as affordable as possible, preferably not require specialised equipment that is not commonly available within schools, and provide clear learning objectives.

Before continuing several NZ schooling specifics need to be explained, as this work was carried out within the NZ curriculum framework and was influenced by this.

6.2.1 *The School Decile System*

In the next three chapters, there are multiple references to school *deciles*. In NZ, the Ministry of Education uses the school decile system to determine how much Government funding a school receives. It is a scale from 1-10, where decile 1 schools receive the highest amount of funding, while decile 10 schools receive the least. The Ministry of Education describes deciles as “a measure of the socio-economic position of a school’s student community relative to other schools throughout the country. For example, decile 1 schools are the 10% of schools with the highest proportion of students from low socio-economic communities, whereas decile 10 schools are the 10% of schools with the lowest proportion of these students”.³ Lower decile schools tend to have a higher proportion of students with behavioural problems, and who live in poverty.

³<https://www.education.govt.nz/school/running-a-school/resourcing/operational-funding/school-decile-ratings/>

6.2.2 *Priority Learners*

The NZ Education system identifies a specific group of students as ‘priority learners’. Priority learners are defined, by the Education Review Office (ERO⁴), as “groups of students who have been identified as historically not experiencing success in the New Zealand schooling system. These include many Māori and Pacific learners, those from low socio-economic backgrounds, and students with special education needs.” [73]

It is crucial that as new subjects are introduced to schools they are taught in a way which gives all students an equal opportunity to learn and succeed, and in NZ special attention needs to be paid to the needs of priority learners. Because of this, throughout the studies I conducted I aimed to gather demographic information on students’ ethnicity and special education needs, and the decile of each school. This was not possible for every school and class involved, due to privacy concerns, the types of information schools collected on their students, and the challenge of collecting permission slips from students.

In circumstances when it was possible to collect data on students’ ethnicity, I investigated whether there were any significant differences between results and observations of Māori and Pacifica students and students of other ethnicities. This covered results in assessments, and also qualitative information on students’ enjoyment, engagement, and attitudes in class. Data showing which students had special education needs was not collected, but in the interviews conducted in 2015-2017, teachers sometimes discussed students with these needs. The decile of each school was recorded, and comparisons between schools of different deciles were made, when appropriate.

6.3 *Resources and training used throughout the studies*

The CS and programming topics I focused on in these studies were the same as those outlined in the ‘Data Representation’, ‘Algorithms’, and ‘Programming’ strands of the proposed curriculum discussed in Chapter 4, section 4.4. These topics, and the specific concepts these covered are shown in table 6.1. These topics dictated the choice of resources and training methods provided during these studies.

This table shows these concepts mapped to specific levels (indicating age groups) of the NZ curriculum. These age groups were guidelines rather than absolute divisions, and teachers were free to teach concepts from different levels to their students’ age group.

⁴ <http://www.ero.govt.nz/>

	Algorithms	Data Representation	Programming
Year 1-3 (5-7 years old) and up	<p>Understand what algorithms are and follow an algorithm.</p> <p>Decompose problems into steps, as a group, with teacher assistance, and/or individually</p>	<p>How 0's and 1's can represent numbers.</p> <p>Binary numbers up to 4 bits.</p>	<p>Programming with 'Turtle' instructions (e.g. basic directional instructions), Unplugged and/or with robotic toys (e.g. BeeBots).</p> <p>Age-appropriate block-based/visual programming (e.g. Scratch Junior) with simple sequencing.</p> <p>Testing and debugging.</p>
Year 3-5 (7-9 years old) and up	<p>Understand that algorithms have to be made up of unambiguous steps, and that we can implement them in programming languages so that a computer/digital device can follow them.</p> <p>Recognise that there can be multiple algorithms for the same problem (e.g. search, sorting).</p> <p>Correct errors in algorithms and/or errors made when following an algorithm.</p>	<p>How two different symbols/states can represent different types of information (e.g. binary numbers).</p> <p>Binary numbers up to 5 bits, or more.</p> <p>Simple binary representation of numbers, text, and images.</p> <p>Understand that digital devices/computers represent information using digits.</p> <p>Error detection and correction algorithms (e.g. parity cards, barcodes)</p>	<p>Block-based/visual programming (e.g. Scratch), including simple iteration, user input and output.</p> <p>Recognise that we have to use specific instructions that computers can understand to program them.</p>
Year 5-8 (9-12 years old)	<p>Recognise that some algorithms (for solving the same/similar problems) are better than others.</p> <p>Comparing and evaluating different algorithms for the same problem.</p> <p>Create an algorithm to solve a problem, by decomposing the problem into steps.</p>	<p>Simple binary representation of numbers, text, and sound.</p> <p>Image representation using pixels.</p>	<p>Block-based/visual programming, including simple iteration, user input and output, variables, selection.</p> <p>Designing a program for a specific purpose.</p> <p>Reusing solutions/parts of programs to solve similar problems.</p>

Table 6.1: The concepts that were used for teaching throughout the 2014 to 2017 studies.

This proposed curriculum was not written until after the Pilot study (Chapter 7) was completed, but the concepts I focused on during that study matched those in the proposed curriculum. As the new NZ DT curriculum was published after the completion of these studies, it was not used for any of them and not available to participants. The topics and concepts covered in my studies however essentially match what is now included in the CT strand of the DT curriculum.

Computational Thinking is not mentioned explicitly in these topics and was not part of the resources and training content provided throughout the studies. These instead covered only CS and programming; there were two reasons for this. During the time the pilot study was conducted, CT was not yet a focus of my research. During the following two studies I did investigate the development of CT skills, but instead of teaching it explicitly I intended to investigate whether these skills were developed by the CS and programming lessons.

Throughout the first two studies, reported on in chapters 7 and 8, I was continually seeking and building on feedback from participants on the training and resources provided. I revised the training and choice of resources used throughout these studies. This feedback also impacted the material taught at the Computer Science for Primary Schools (CS4PS) workshops provided by the CSERG during the time the second study was conducted. These workshops are described in section 6.3.4.

Throughout all three studies, I was available to answer teachers questions on the content they were teaching and to provide extra training and support if they requested it. In the first study I team-taught four lessons (each a different topic with the same class) with the participating teacher, and one lesson each with four different teachers and their classes during the second study, at their request. During the second study Wendy, a member of the CSERG, worked with five other teachers and supported them through one-on-one training and in class support.

6.3.1 Differences between participants training in the second study

During the second study, the Open Teachers study covered in Chapter 8, the amount of training and support each participant received varied substantially. This was due to the freedom teachers had in choosing when to join and leave the study, what resources they would use, what selection of topics they would teach, and what types of support they wanted. Some participants attended one or more of the CS4PS workshops, described in section 6.3.4, which were run by the CSERG. Others did not attend any of these workshops and worked

independently, using the online resources provided or other resources they sought out. Some received ongoing training and support, through one-on-one tutorials, team-teaching sessions where I joined them for a class, and frequent email contact. Some participants worked together with colleagues, within and outside of their schools, who were also taking part in the studies, to support each other in their learning and teaching. One-on-one training, email support, and team-teaching support were offered to all participants. Participants were provided with as much support as they requested so that a lack of support would not be a major factor in the results of these studies. This extra support was only requested by a small number of teachers.

These differences are further discussed in Chapter 8.

6.3.2 Computer Science Unplugged

The following Computer Science Unplugged activities and materials were used for teaching Algorithms, Data representation, and some of the Programming concepts.

- **Binary numbers**

Using a set of 4 or more ‘binary cards’, (usually white with dots on one side, and entirely black on the other, but variations can be used) students learn to represent base 10 numbers in binary, and how to count in binary. They observe patterns, such as each card having twice as many dots as the previous, and how adding another binary card (or bit) allows you to represent twice as many numbers. This can be extended by using different things to represent 0 and 1, for example ‘on’ and ‘off’, up and down arrows, koru patterns pointing in two different directions, or high and low pitched sounds.

- **Binary representation of text**

By assigning the numbers 1-26 to the letters of the alphabet, students learn to represent English text in binary. This can be extended by adding more letters or symbols to the set of numbers (e.g. lower case and capitals, or macron letters for writing in Te Reo Māori)

- **Image representation**

Students are given a set of numbers and a grid, and must use these numbers to figure out which boxes on the grid to colour in. If the grid is correctly filled out then a picture, which was encoded in the numbers, will be revealed.

- **Kid-bots (programming Unplugged)**

Students take turns being a ‘robot’ or a programmer. The ‘robot’ can only understand a small set of specific instructions, and the programmer’s task is to write a program using these instructions, to have the robot accomplish a task. The most frequent way this is used is with the robot standing on a grid, and they are only able to understand the instructions “turn left” (90 degree turn on the spot), “turn right”, and “move forward” (take one step forward, in the direction you are facing).

- **The Parity Magic Trick**

Using a set of ‘parity cards’ to perform a ‘magic trick’. Parity cards are square cards with a different colour on each side (e.g. black and white). The person performing the trick seems to always know when a card has been flipped over without watching, but they are in fact using an error detection and correction algorithm.

- **The Barcode Magic Trick**

After being told all but the last digit on a barcode, the person performing the trick can ‘magically’ tell what the last digit is. The person performing the trick is actually using an error detection algorithm, as the last digit on the barcode acts as a check digit.

- **Sorting Networks**

Using a ‘Sorting network’ on the ground (this looks similar to a flow chart), students are each given a different number and follow the arrows on the sorting network, comparing their numbers to other students as they go along, and without intentionally doing so they find they are in a sorted order when they arrive at the other end of the sorting network. This activity can be done with anything that can be put into an order, not just numbers. For example it can be done with letters or words. The activity is intended to teach the concept of concurrency, and also introduces the idea that a sorting algorithm can be constructed using simple comparisons between two things.

- **Searching Algorithms**

Students experiment with using the searching algorithms: linear search (also called sequential search), binary search, and (as an optional extension) searching a hash table. There are several variations on this. The two that were most commonly used in these studies were: searching through a line of cups with ping-pong balls in them, for

the ball with a specific number on it; and through a battleships game, where students search for each other's ship based on the number on that ship.

- **Sorting Algorithms**

Using a set of weights and a balance scale, students use two different algorithms to sort the weights into order from lightest to heaviest, by comparing only two at a time, and count how many comparisons are made for different algorithms. This can be extended by trying out more sorting algorithms.

- **The Muddy City**

An activity for illustrating graph problems (minimal spanning trees and networks). Students are given a map of a city with paved paths leading between different buildings, and have to choose the paths (by colouring them in), which together allow you to visit every place on the map, whilst using the least number of pavers. This can lead into discussions of optimisation, algorithms for finding the best set of paths, and variations involving different requirements for optimising the network.

The full descriptions of these activities can be found through the CS Unplugged website⁵, and the Classic CS Unplugged website⁶. Teachers were able to pick and choose which of these they used (other than several activities that were required lessons in the 2017 study). These activities will be referred to frequently throughout the remainder of this thesis.

During my research, the UC Computer Science Education Research Group began redeveloping and extending a subset of the original CS Unplugged activities. The aim of this was to create a version that could serve as a full classroom programme for use in primary schools. I was also involved with this redevelopment. The new versions were made available to participants in 2016 through a 'sneak peek' website, and this new version of CS Unplugged was officially launched in February 2018. This meant that during my research the CS Unplugged resources I was able to provide teachers with changed over time. A small number of teachers participating in the second study and all teachers in the third study used the updated versions of the resources.

The activities which were redeveloped and released were: Binary numbers, Binary representation of text, Parity Magic Trick, and Sorting Networks. The Kid-bots activity was

⁵ <https://csunplugged.org/>

⁶ <https://classic.csunplugged.org/>

developed for the new version of CS Unplugged, and so was not available through Classic CS Unplugged and not used by participants until 2016. The remaining activities were not redeveloped during this research and available only through Classic CS Unplugged.

In the first study, when I worked with just one teacher, they used the CS Unplugged book⁷ and online resources for her classes, and I went through each of the CS Unplugged activities with her once before she taught them. I taught three of these activities (binary numbers, the parity magic trick, and searching algorithms) once alongside the teacher, before she proceeded to teach them with more classes. She was free to choose which of the activities to teach.

In the second study, teachers were given the freedom to choose which CS Unplugged lessons and activities they would teach. All of these participants had access to the CS Unplugged lessons and activities, whether it was through the Classic CS Unplugged or the new CS Unplugged website. As mentioned previously in section 6.3.1, participants in this study received varying amounts of training in CS Unplugged, based on the amount of support they requested and how many of the CS4PS workshops they attended.

For the final study, I conducted one-day workshops at each of the three schools I worked with, to train the teachers on using the CS Unplugged activities. Teachers then had access to the CS Unplugged online resources and could contact me if they had questions about the content. During this study, I did not do any CS Unplugged team-teaching.

The specific CS Unplugged lessons and activities used by the teachers are documented in each of the chapters on these studies.

6.3.3 Programming resources

Throughout the first two studies, reported on in Chapters 7 and 8, teachers were given full autonomy to use whatever programming resources and languages they wanted. In the final study, reported on in chapter 9, I instead assigned specific resources for teachers to use.

During the first study, the teacher chose to use the programming language Scratch⁸, and I provided her with resources and one-on-one tutoring in programming using Scratch.

⁷ <https://classic.csunplugged.org/books/>

⁸ <https://scratch.mit.edu>

Programming in the second study

Training in using the visual programming languages Scratch and ScratchJr⁹, and the Bee-Bot¹⁰ floor robots, was provided through the CS4PS workshops. Information on what Bee-Bots are is given at the end of this section. Teachers were also offered one-on-one tutoring in using these programming languages but, to my knowledge, only one teacher requested and received this. Several participants took part in a ‘Code Club for teachers’, run by Wendy, where they worked together on the Code Club Aotearoa¹¹ Scratch projects and discussed how they could incorporate these into their classroom programmes. Teachers were free to choose the programming languages they used. Scratch and ScratchJr were the most common languages used.

The Bee-Bots were mostly used with grids on the floor, and teachers had students program the robots to navigate the grid in a specific way to achieve a goal. These lessons included navigating mazes and obstacles, spelling out words with letters printed on the grid, and performing basic maths with numbers on the grid. The Bee-Bots proved extremely popular with both students and teachers, and a group of teachers decided to organise an inter-school competition with these, named the “Buzz-off”. This was independent of the work I did but involved the UC Computer Science Education Research Group. This is a testament to how effective teachers found the Bee-Bots to be as learning tools. The Bee-Bots were used to teach Algorithms and Programming.

Along with this training, we recommended several online resources. The main programming resources I directed teachers towards were the projects published by Code Club Aotearoa, and these were the most commonly used resources, particularly the Scratch ones. I also suggested several ‘learn to code’ websites, such as Codecademy¹², CodeAvengers¹³, and CodeCombat¹⁴. To my knowledge, these were only used by two teachers, and only by the more advanced students in their classes who required an extra challenge.

Several teachers used other resources they found online or were recommended by their peers. The resources that were reported were the Hour of Code activities and Code Stu-

⁹ <https://www.scratchjr.org/>

¹⁰ <https://www.bee-bot.us/>

¹¹ <https://codeclub.nz/>

¹² <https://www.codecademy.com/>

¹³ <https://www.codeavengers.com/>

¹⁴ <https://codecombat.com>

dio courses available from Code.org, and the “Code It How To Teach Programming Using Scratch” teacher’s handbook[8]. One teacher said they used several Scratch programming tutorials written by other teachers, that they found on the *Computing at School* [44] forums, but did not provide further information. Other resources I was not aware of may have also been used. The more confident teachers frequently adapted resources to suit other topics and projects taking place in their classroom.

The CS Unplugged activity ‘Kid-bots’ can also be considered to teach both Algorithms and Programming, as the activity involves the same concepts as using the Bee-Bot floor robots. The main difference is that in Kid-bots the instructions are interpreted and carried out by a human, rather than a robot or other digital device.

Bee-Bots

Bee-Bots are educational robots that have the appearance of a bumblebee. They have a set of buttons on their back, shown in Figure 6.1, that can be used to program them with a set of movement instructions. They can only perform simple sequences of instructions, and looping and conditional instructions are not available. Several similar robotic toys have been developed, such as the Code and Go Programmable Robotic Mouse¹⁵.

Programming in the third study

In the final study I conducted, the Intervention study discussed in Chapter 9, teachers were assigned one programming resource, in an attempt to maintain as much consistency between different classes as possible. Teachers were asked to use CodeAvengers¹⁶ (and the KidBots CS Unplugged activity, which can be considered to teach programming concepts) only. However, due to technical issues that arose with using CodeAvengers, partway through the study teachers were given permission to use the Code Club Aotearoa Scratch programming resources.

6.3.4 Computer Science for Primary Schools (CS4PS)

Along with direct support from myself and another member of the CSERG, many participants attended one or two of the ‘Computer Science for Primary School’ (CS4PS) workshops,

¹⁵ <https://www.learningresources.com/product/learning+essentials--8482-+programmable+robot+mouse.do>

¹⁶ <https://www.codeavengers.com/>

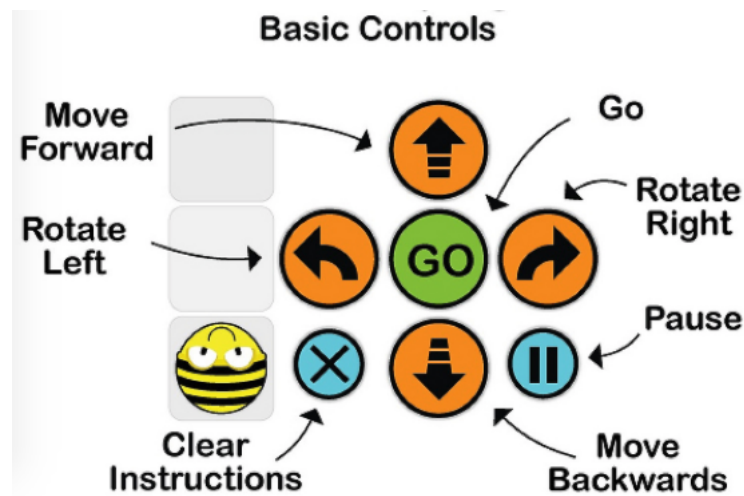


Figure 6.1: The Bee-Bot buttons, are used to program them to move forwards, backwards, or to rotate 90 degrees in place. From the Bee-bot iOS app <https://itunes.apple.com/nz/app/bee-bot/id500131639>

which took place between 2016 and 2017. These workshops took place over one or two days and were free for teachers to attend. They were run by the UC Computer Science Education Research Group, and volunteer teachers who had prior experience with teaching these topics. These teachers were all people we had previously worked with. At these workshops, teachers participated in discussions on CT, shared their experiences with one another, and sessions were run on teaching with CS Unplugged, programming in various block-based languages, and programming with Bee-bots. During these workshops, I provided support for my colleagues and observed sessions, but did not provide any training myself. Teachers were recruited during these workshops for the second study conducted, the Open Teachers study discussed in chapter 8. They were informed that the goal of this study was to gather feedback on the resources they used and to investigate the impact teaching these topics had on their students.

6.3.5 Teacher Collaboration

In all cases where several teachers from the same school were taking part in these projects, they described collaborating and supporting each other throughout. Collaborative relationships also organically grew between several teachers at different schools, who had met through CS4PS workshops, or one of the many events related to Digital Technologies that

were run over the years by our research group and others. Some of these teachers went on to run their own professional development workshops and events for other teachers, within their schools and outside of them. Several of the participants in the second study attended the ‘Code Club for teachers’ in 2015. These relationships were cited by many teachers as crucial to their success in integrating these topics.

6.4 *Data collection and analysis methodology*

This research was conducted in New Zealand primary school (ages 5-13 years) classes. The majority of these classes took place in usual class time; involved an entire class of students, rather than a specially selected group; and were taught by a generalist primary school teacher who had not previously taught CS and programming. In the second study I conducted (covered in Chapter 8), there were a small number of cases that were exceptions to this, which are each noted.

I took a pragmatist approach to this research [168, 166], focussing on the real-world context and effect of this work. I did this because the goals of my research centred on supporting action in CS and CT education. Any findings from this work were intended to be of practical and relevant use. Pragmatic research is by its nature reactive. The ability to react and adjust to shifting circumstances was crucial for my work, as curriculum in this area was continually developing throughout my research, both in NZ and abroad. The training and support required by educators expanded, and several of the CS education groups I have been involved with were subcontracted by the Ministry of Education, which impacted what parts of my work had the most practical value. I choose this research approach because I needed flexibility in how I conducted my studies and what information I needed, and was able, to collect.

The studies I conducted each took a mixed methods and classroom-based research approach. Mixed methods approaches are useful for research in educational settings because of the complexities of researching within a school environment. When working with multiple people who need to prioritise their teaching and students over participating in my research, and the constantly evolving and dynamic environment of a classroom, there is no way to have control of all variables involved in the research. A mixed methods approach assists in the triangulation of results based on multiple data sources. With no established CT assessment tools, it was also important that I supplemented the quantitative data from tests with qualitative information. The types of data collected, and the data collection methods used,

varied between each study, and these are described in more detail within each chapter.

6.4.1 Assessment with Bebras

As previously stated in section 3.4.1 of Chapter 3, the Bebras challenge¹⁷ is an “International Challenge on Informatics and Computational Thinking”. It was used in two of the studies I conducted to try and measure students’ CT skills. Along with its favourable reputation among those who had used it, and the consensus that the majority of tasks in the challenge require CT skills to solve [98], I chose to use this challenge for the following reasons. Firstly, the challenge does not require prior experience with CS or programming. This was useful as it meant students would, in principle, be assessed purely on their CT skills rather than their achievement being based on prior experience. This was particularly important for pre and post-testing students as I was aiming to assess their development of CT skills, and not just their knowledge of CS and programming. Secondly, it was accessible online and did not require schools to install specific software or deal with troubleshooting software. Thirdly, there are different sets of questions available for different age groups of students, meaning it was potentially useable for multiple primary school year groups. As discussed in Chapter 3 however, there are several limitations to using the challenge, and there has been criticism of its use and validity. Please refer to section 3.4.1 for more details on these limitations.

The challenge follows a specific marking schedule, where students gain points for correct answers and lose points for incorrect ones. Points are not lost for unanswered questions. All students start with a base score of 45 points to avoid obtaining a negative mark. Each challenge has a minimum possible score of 0, and a maximum of 180. A score of 0 indicates that a student attempted every question, but answered all incorrectly.

Each task is assigned a level of difficulty: easy, medium, or hard. The difficulty of a question dictates the number of points that are gained or lost when that question is marked. A correct answer to an easy question earns 6 points, while a wrong answer loses 2 points. Medium questions earn 9 points or lose 3, while hard questions earn 12 points or lose 4. Levels are assigned to each question by the Bebras task developers, at their discretion, when they develop new tasks at the annual international Bebras workshop. These differences in scoring for each task can be beneficial as it gives a larger and potentially more informative spectrum of possible results, and it gives recognition for solving particularly difficult questions. However, it can also introduce problems. As discussed in Chapter 3, the

¹⁷ <https://www.bebbras.org/>

assigned difficulty of a task is at times inappropriate and does not reflect the actual difficulty level for students in practice. For example, a question marked as medium may have been much more challenging than the task developers intended, and fewer students answered it correctly than some of the questions marked as hard, or vice versa. In the case of the Pilot study, covered in Chapter 7, I had access to each student's final score, but not to their marks for each question. This prevented me from assessing whether the assigned difficulty levels of the tasks matched the actual difficulty, based on how many students answered each correctly or incorrectly. It also made it impossible to tell the difference between a student who got several questions correct and several incorrect, and a student who did not answer several questions.

In the final study I conducted, I did have access to the answers to each task and so I was able to examine the difficulty level of each task. As several of the assigned difficulties were found to be substantially different from the actual difficulty level, I re-assigned levels to these tasks and re-marked students results according to this more accurate rubric. However, to further complicate the issue of task difficulties, the actual difficulty of a task is not always a result of the informatics concepts or skills that need to be exercised. It can also be influenced by the language used, and length of the question, as this is not always suitable for the target age level. This issue arose during the third study I conducted. This issue and its implications are discussed further in Chapter 9.

6.4.2 Qualitative data

All survey questions, questions in the feedback form, and the planned format and topics to be covered in interviews were reviewed and approved by the University of Canterbury Educational Research Human Ethics Committee.¹⁸ See Appendix C.

Pilot Study

During study 1, the Pilot study, I collected qualitative information by taking notes on my meetings with the participating teacher, the observations we each made during and after the lessons I was present for, and on any online communications we had. Our meetings were not audio recorded or transcribed.

¹⁸ <https://www.canterbury.ac.nz/study/ethics/educational-research-human-ethics-committee/>

Interviews in the Open Teachers Study and the Intervention Study

In the subsequent studies, I conducted interviews with participating teachers. I recorded the audio from these and transcribed them. For study two, the Open Teachers study, these interviews were voluntary. All participants were invited to volunteer, regardless of when they joined the study or how long they had been a part of it. The one teacher who participated in the first study also participated in the second study and was interviewed. In study three, these interviews were a requirement of participating in the study and all participants consented to be interviewed. During the second study, the interviews lasted between approximately 30 to 75 minutes, depending on how much the interviewee wished to discuss. The interviews conducted for the third study were shorter, between 20 to 40 minutes, as participants were assigned lessons to do with their classes and so did not have any extra activities or curriculum integrations to report on. Around half of the interviews in the second study and the majority of interviews in the third study were conducted in person. The remaining interviews took place over the phone.

These were semi-structured interviews and so a script of specific questions was not used. Instead an interview guide, based on the general topics for discussion, was used. This allowed for increased flexibility in the interview process. Because of this, I was also able to explore topics participants brought up that I had not predicted. This approach generated insightful, and in some cases surprising, information. However, in some cases, the variation in topics limited the potential for comparisons of particular observations between participants.

The topics discussed in the interviews were:

- The participant's background in teaching, and how they came to be involved in this programme.
- The general culture of their school and/or teaching environment.
- The students in their class(es). For example, whether the students were particularly high or low achieving, motivated learners or disengaged, if there were particular behavioural issues within the class, etc.
- Which lessons they did and their observations about the lessons: Success, benefits, positive and negative impacts, impact on students general learning and their problem solving skills.

- The impacts being a part of the programme had on them personally, what their feelings about the programme were, and what challenges they faced.
- What types of support and resources were most helpful to them.

Care was taken to avoid asking leading questions, for example asking “did you observe any positive or negative impacts on students?”, rather than “what positive impacts on students did you observe?”.

Additional data in the Open Teachers Study

Participants were asked to fill out a survey when they first joined this study, which asked about their motivations for joining the study and teaching this material, their feelings about the prospect of teaching this, and their previous experience with these subjects. These were all text entry questions.

Throughout the study, participants were also asked to submit feedback on the lessons they conducted through an online form. This form included three optional text entry questions and two required multiple-choice questions with an ‘other’ option where text could be entered.

The qualitative information submitted through the survey and the feedback form was analysed alongside the interview transcripts from this study.

Analysis

I used a thematic analysis approach when analysing the qualitative data collected in each study. The analysis of the second study was completed before the third study began. Thematic analysis is a widely used method for analysing qualitative data and is used to identify patterns and meanings across a data set related to the research questions being investigated. To identify these patterns and meanings a process of ‘coding’ data is used (not to be confused with ‘coding’ in the context of programming). This tags sections of text with a code (generally a key-word or potential theme). It is an iterative process in which the researcher goes through phases of becoming familiar with the data set; coding the data; revising and validating the codes used, which involves reviewing and (when necessary) re-coding the data; and finally taking the codes and coded data to consolidate and describe the themes found in the data.

The aim of collecting and analysing this information was to identify information related to Computational Thinking, and to investigate if other themes or patterns related to students' and teachers' experiences emerged. I used a deductive approach when investigating CT skills and themes in this data, attempting to identify data that fitted into the CT framework I was using. As discussed in Chapter 3, section 3.5, I used the Barefoot CT definition in my analysis of interview transcripts and participants feedback. While this definition does not fully agree with my own view of CT, it was used for pragmatic reasons, as the approaches it includes have close links with core parts of the NZ curriculum. These links are discussed in Chapters 8 and 9. I then employed an inductive approach to my analysis, attempting to identify themes or patterns in the data that were not based on a preconceived framework or set of themes. It should be noted that in inductive analysis it is never possible to be certain that a researcher's preconceived theories about the data have not impacted the analysis.

6.5 Limitations and threats to validity

There were a number of limitations and threats to validity for the methodology used in these studies that need to be noted. The main limiting factor throughout all of my studies was that my involvement in the NZ curriculum redevelopment, and the needs of teachers (both participants in my study and others my research group were working with) took priority over my research. Balancing the needs of individual participants in my studies, and the wider community of NZ primary school teachers, with the repeatability, rigour, and goals of my research was an ongoing challenge.

More specific limitations of my work are discussed below.

6.5.1 Pragmatism and mixed methods research

Mixed methods approaches often become time consuming, and require complex research designs. Pragmatic research also includes these challenges, as it is reactive and the research being carried out must respond to changes in the real-world context. Changing circumstances mean findings and conclusions can become meaningless, or no longer applicable to the current real-world situation, at different points in time. The reactive nature of pragmatic research also greatly decreases its repeatability. The focus on generating results that are of practical value, and in my case the reliance on participants' personal experiences, can impede repeatability and rigour. This is a common problem in education research and other research involving human behaviour. Despite these disadvantages of mixed methods and

pragmatic research methodologies, their advantages made them well suited to my research.

6.5.2 Interviews and interview analysis

During these studies, I was closely involved with the participants and worked alongside them continuously. This familiarity was highly beneficial in providing participants with adequate support, and in increasing their engagement with the study so they were more willing to provide feedback and be interviewed.

However, this also introduces several threats to the validity of these studies. The first of these is acquiescence bias; participants may want or feel a need to please the researcher, and so give overly positive reports of their observations and experiences, compared to the reports they may have given if they had not been as closely involved with the researcher. This risk could have been reduced by having a neutral third party conduct the interviews. On the other hand, having a third party who was not familiar with research would have likely limited the depth of information gathered in the interviews. There is also an inherent threat to the validity of conclusions that are based on teachers' opinions and impressions [92]. It introduces the risk of participants giving overly positive reports of their students' learning. Examining a teacher's impression of their students' learning can provide valuable information, but can not give complete access to students' actual knowledge.

During interviews, care was taken to avoid leading questions that could bias teachers' answers. However, my reactions to participants answers may have introduced bias into their subsequent answers. For example, there were several cases in the second study where I responded to participants with statements such as "that's really exciting" or "that's good" when they gave answers in a particularly enthusiastic and excited fashion, or which were interesting to me. I had attempted to remain impartial while interviewing, but during transcription of interviews in the second study, I observed several cases where I gave positive reactions to the interviewees' statements.

I was the only researcher who analysed the interviews, and so there was no secondary thematic coder that I could compare my results with. The primary reason for this was there were no other people available to conduct this lengthy process of analysis. It was therefore not possible to calculate inter-rater reliability of the coding. As previously mentioned, the use of inductive thematic analysis means the impact of a researcher's preconceived theories about the data may bias their analysis, and without the ability to check inter-rater reliability the potential degree of bias in the analysis could not be examined.

6.5.3 *Assessment with the Bebras challenge*

As was discussed in Chapter 3, section 3.4.1, the reliability of the results of the Bebras challenge as a measure of CT skills has some limitations. There is variation in the opinion of researchers on whether the Bebras challenge is a suitable method of CT assessment, particularly because the reading difficulty of the challenge and emphasis on creating narratives within the questions can distract from the CT concepts the challenge is aiming to assess [150, 184]. In the third study (the intervention study) I attempted to control for the reading difficulty of the challenge by collecting data on students' reading ability, along with their test results. The accuracy of the difficulties assigned to questions has been found to be unreliable [23, 184]. I attempted to mitigate the effects of this by reassigning the difficulties of questions in the tests I used for the third study but was unable to do this in the first study (the pilot study) as I only had access to students' final scores.

6.6 *Additional considerations*

6.6.1 *Impacts of participant's location*

All participants were teaching in NZ, with the majority based in the Canterbury region. The higher concentration of participants in Canterbury was a result of the CSERG, CS4PS workshops, and myself being based in Christchurch, Canterbury. The potential influences of this regional bias were not investigated. Christchurch is a fairly typical NZ urban region. However, several participants discussed the impacts of the 2011 Christchurch earthquake¹⁹ during their interviews. These discussions focused mainly on the effect of the earthquake on their student's well being and behaviour problems.

The large detrimental impact of the earthquake on the mental health of young people in Canterbury has been heavily documented. The frequency of behaviour problems in students has been increasing since 2011, and primary school students in Christchurch have approximately double the normal rates of anxiety and PTSD symptoms for their age group [118]. The trauma of the earthquake has also had an impact on children born after it occurred, meaning even the youngest students in these studies were potentially affected [32, 108]. The 2011 earthquake, student behavioural problems, and the impact of this study on these problems became a common theme in the interviews.

It is also worth noting the differences in population proportions of Māori, Pacifica, and

¹⁹https://en.wikipedia.org/wiki/2011_Christchurch_earthquake

European identifying people in the South and North Islands of NZ. The Canterbury region, along with most regions in the South Island, has a relatively low percentage of people who identify as Māori or Pacifica, and a relatively high percentage of people who identify as European, when compared with the majority of regions in the North Island. In NZ there is a significant difference between Māori and Pacifica student achievement rates when compared with students of other ethnicities. There are a large range of factors contributing to this, including structural problems within the NZ education system. The results of this research may, therefore, be less representative of teachers and students in the North Island of NZ. However, the demographics of the schools I worked with varied greatly so this potential for bias may have been decreased.

6.6.2 Discussions on gender

In all following chapters, discussions and presented results that concern gender only discuss cis-gender participants (students and teachers). This is because in each study there were, to my knowledge and in the data collected, no gender diverse participants in the 2015-2016 open teachers study, or the 2017 intervention study. In the 2014 pilot study, 4 of the students responded ‘other’ when asked what their gender was. As this was too small a sample size to make any meaningful generalisations about, and none of the students’ results placed them as outliers in the data set, no conclusions were drawn based on students responding as ‘other’ and being gender diverse.

Because of this, in discussions of students only girls and boys are referred to, and only women and men in discussions of adults.

6.6.3 Discussions on ethnicity

In all following chapters, discussions and presented results that concern students’ ethnicities focus on students who are Māori or Pacifica, and students who are neither Māori or Pacifica. In these discussions, I have grouped together students whose includes either Māori or Pacifica, or both. This also includes students who identified as additional ethnicities, for example a student who identified themselves as Māori and ‘New Zealand European (Pākehā)’ is included in the group Māori/Pacifica. Throughout the following chapters, this group of students will be referred to as Māori/Pacifica. It is important to note that these two ethnicities are not the same, and Pacifica (also referred to as Pacific Peoples and Pasifika) is a collective name for a number of Pacific Island ethnicities, including Samoan, Tongan,

Niuean, and Cook Island Māori.

These ethnicities have been grouped together in discussions on ethnicity as both groups face similar inequality issues in school education, and are both similarly unrepresented in the computing industry and computing courses. These issues were previously discussed in Chapter 2.

Chapter VII

Pilot Study

This chapter covers the Pilot study I conducted in 2014, with an intermediate school teacher who was completely new to the topics of CS and programming. First, the circumstances which led to this study, and the aims the teacher and I had for introducing these topics are described. The structure and content of the CS and programming classes are discussed, followed by the methods of assessing the success of these classes. The results of this programme and the implications these had for future studies are then discussed.

7.1 Context

In 2013 Angela¹ was teaching technology at an intermediate school and was dissatisfied with the classroom program being used for ICT. She observed that many students were disengaged, bored by the content, and were not leaving the class feeling that they had achieved something. They spent most of their lessons working through web-quests,² which are online enquiry oriented lessons in which students search for information on the web. This was problematic, as these required large amounts of reading and thus shut out many students with lower reading ages. In general, these tasks were not motivating the students.

In 2013, Angela was given the opportunity to take over the classes in the school's computer lab. This enabled her to change the content, and the way students were learning in ICT. She spent the year focussing on teaching digital media and encouraging students to create their own media. She felt this was successful, but wanted to take things further.

In early 2014, serendipity had it that Angela heard about our intention to work with primary school teachers, trialling CS and programming in classes. She volunteered to work with us. Over the next school year, I worked with Angela to implement a trial curriculum and conduct a pilot study at her school. During this time, I trained Angela in teaching

¹ Name has been changed.

² <https://en.wikipedia.org/wiki/WebQuest>

CS using CS Unplugged activities, and programming in Scratch. At times I joined her to team-teach classes. Throughout the year I visited Angela at her school to go through CS and programming lesson plans, and discuss her experiences with the material. I visited her weekly for the first school term, and approximately biweekly as the year went on and she required less support. In each of our meetings, I took notes on her recent in-class experiences with the content, what she was struggling with and wanted to learn, and what content I taught her during our meeting. Throughout this project, I took on the role of teacher to Angela and was simultaneously investigating whether the resources she was supported with were effective for teaching and her students learning.

Angela and I attempted to assess students' learning, and investigate the impact of these lessons on them. We did this through quizzes, the online Bebras challenge, and an attitude survey that students completed. The results from these, and Angela's ongoing observations and feedback throughout the year, provided invaluable insight into the impacts and challenges of teaching CS, CT, and programming, in an intermediate school environment. The findings from this study informed the design of the larger-scale study conducted through 2015-2016, which is covered in Chapter 8. They also had a large impact on the official discussions surrounding the NZ curriculum development in 2014, and the following years.

Angela remained involved in this work over 2015-2016 and was a participant in the study covered in Chapter 8. At the end of 2016, she was interviewed about her experiences over all three years. The contents of this chapter at times draw on information from the interview, which relates to this pilot study.

7.2 Aims

This pilot study focused on exploration and discovery. The intent was to investigate the general impact on students and teacher, e.g. on their attitudes to the topics, engagement in the lessons, and the teacher's professional opinion on the students' learning. Angela's goals for the study were as important to its planning and conduct as mine.

In February 2014, I had my first meeting with Angela. She had been teaching for over 25 years, the past 18 being in intermediate schools. She was a generalist teacher, with no previous experience or exposure to CS or programming. We discussed why she wanted to learn about, and teach these new topics, and what she wanted to achieve that year. Her goals for students were:

- They would be engaged in the content and enjoy the classes.

- Every student would be able to achieve something in these classes and leave with a sense of personal achievement.
- They would have some control over their own work and would be encouraged to be creative.
- They would have the opportunity to create something that they were able to share outside of the classroom with their family, friends, and whānau.
- They would have opportunities to work together and collaborate.
- They would be encouraged to assist each other in solving problems and share their learning with each other.
- Extension activities could be provided for those students who needed them, but the main course content would be suitable for all students.
- At the end of the year, students would be able to get an indication of how well they did and whether they understood what they had learnt. This would not necessarily be through some form of assessment, but this was an option.

Angela also wanted to be able to keep up with the majority of her students throughout the year and was concerned that she may not be able to do this. She expected that some high achieving students would excel and move quickly through the content, and did not mind if she could not keep up with them, but it was important to her that she kept pace with the rest of her students. She did not want students to be held back if she was unable to progress. She initially did not have high levels of confidence in her ability to teach CS and programming. However, she was highly confident in her computer use skills, in incorporating E-learning into the classroom, and was not afraid to experiment with new software. While these skills are not directly related to CS and programming, this removed the potential extra barrier of a lack of confidence with technology, which some late career teachers struggle with.

My aims for this study were to observe and investigate the general impacts of introducing this material to the classroom. The specific impacts I focused on evolved slightly over the course of the study. I progressed from investigating students specific CS knowledge and skills, to investigating their CT skills as well. This was in response to the growing catalogue of literature on CT, and that this was impacting opinions among CS educators and curriculum

influencers. Development of CT skills was increasingly being viewed as a more important outcome for students, than simply gaining CS knowledge. Learning CS was becoming seen as a means to learn CT, rather than CS being the core focus.

Taking into account my initial goals, and the changes in these during the study, the questions I aimed to investigate were:

1. If a teacher without previous Computer Science or programming knowledge could be supported and enabled to integrate Computer Science and programming into their classroom curriculum, through the use of Computer Science Unplugged resources and tutoring in block-based programming (supports RQ1).
2. Whether the teacher developed Computational Thinking skills, or some understanding of these, through learning and teaching Computer Science and programming (supports RQ1).
3. Whether students in a typical classroom could be taught Computer Science and programming concepts, through their teacher's use of CS Unplugged resources, and teaching of block-based programming (supports RQ1.3).
4. Whether students developed Computational Thinking skills, and general problem solving skills, through this process (supports RQ2 and RQ3).
5. How students reacted to and engaged with these lessons (supports RQ2).
6. The degree to which students enjoyed the CS Unplugged and programming lessons, how confident they were in their abilities in programming and the CS Unplugged topics, and if they wanted to continue learning about them (supports RQ4).
7. Whether these observations and results varied for students of different genders, or different ethnicities (supports RQ4).

These questions do not fully match the overall research questions (listed in Chapter 1, section 1.1) I aim to address in this thesis. This is because the process and results of this study, and developments in the NZ curriculum, influenced my subsequent research goals, and informed the research questions I eventually chose. These also contributed to answering several of the research questions, as noted above.

7.3 *Class Structure*

As this was an intermediate school the students ranged from ages 10-13 years (10-12 at the start of the school year, 11-13 by the end) and were in either year 7 or year 8. The ‘computing’ classes with Angela were attended by every class in the school, which totalled over 600 students. Each class spent one 60 minute lesson per week in the computer room. Classes had approximately 20 lessons in total on CS and programming. These lessons were spread out over the school year, though did not take place every week due to school events, timetabling changes, and other ICT topics being taught.

There was also an extension class, who had lessons twice a week. This class received approximately 30-40 lessons in total. The initial sessions with this class served as a type of ‘trial-run’ lesson for Angela and provided her with time to test out the material. I joined the majority of these sessions at the beginning of the year to assist Angela in demonstrating this new material, and answer students’ questions. These sessions gave Angela confidence in teaching the following sessions alone and helped her develop a deeper understanding of the material she was teaching. She remarked during the study that by the end of each week she felt like she was becoming an expert in the topics she had covered. By the middle of the school year, my teaching assistance was very rarely needed.

The classroom environment this study took place in was, what can be described as, a typical school environment. It was a state school, with a socio-economic decile of 5.³ There was a wide range of ability and interest levels across the student body. As a female teacher in the 55-65 year old age group Angela also fell into the typical demographic of primary school teachers, over 80% of whom are women⁴ and 40% are above the age of 50.⁵ A 2012 survey of 91 High School Digital Technologies teachers in New Zealand found that 60% of these teachers were also over the age of 50.

The classroom the lessons took place in was equipped with desktop computers, with enough for one per student. Students generally worked individually on a computer, but frequently collaborated with the other students around them. The classroom was also set up for non-computer work, with a whiteboard, projector, and enough space for activities which did not take place on a computer, such as the CS Unplugged lessons.

To prepare Angela for the lessons, and to support her throughout the year, I met with

³ For an explanation of the decile system please see Chapter 6, section 6.2.1

⁴ https://www.educationcounts.govt.nz/publications/schooling/teacher_census

⁵ <https://www.educationcounts.govt.nz/publications/series/2263/31417/4>

her weekly for the first term, and then roughly biweekly for the remainder of the school year as she began needing less support. In our meetings, we went through the CS Unplugged resources in detail and discussed possible data representation projects and extensions for her to use in class. I also tutored her in programming with Scratch.

7.4 Class content

We initially planned to include ‘Data Representation’, ‘Programming’, and ‘Algorithms’ (as they are described in Chapter 4, section 4.4) in the course content. However, due to limited classroom time, ‘Algorithms’ was not given individual lesson time. Students were exposed to concepts from this topic, as the programming and data representation activities used covered several algorithms, and required algorithmic thinking. However, this did mean that some of the core concepts of the ‘Algorithms’ topic were not covered. Specifically some of those that were highlighted in the proposed NZ curriculum: that there can be many different algorithms which solve the same problem, and that these algorithms can be compared.

The CS Unplugged material on *binary numbers*, *text representation*, *image representation*, and *error detection (parity magic trick)* were used to cover the topic of Data Representation. Each of these activities, particularly binary numbers and error detection, also provided good examples of using algorithms. As Angela became more confident in her conceptual knowledge she began creating activities and challenges for students, based on Data Representation. She discussed these with me before using them. In creating these she consistently displayed an understanding of the topics, and her activities were highly successful. An example of one of these activities was to use different symbols to represent text in binary, rather than using the original black and white cards from CS Unplugged exercise. She had students create art with hidden messages encoded in them, using koru⁶ patterns facing in different directions to represent the two states needed to create binary numbers.

Programming was taught using the educational programming language *Scratch*.⁷ Students used this in all their programming work. Angela and I had both previously observed that when students were shown Scratch and told ‘just have a play and explore’, or given similar instructions, they frequently did one of two things. Some would find this anxiety-inducing as they did not know where to start and become paralysed with uncertainty. On

⁶ The Koru is a spiral shape which is an integral symbol in Māori culture, art, carving, and tattooing, and is a very well known symbol in NZ

⁷ scratch.mit.edu

the other hand, a large proportion of students had a very different reaction and created programs which Angela described as ‘*Sprite-fests*’. These programs tended to contain a small number of short scripts, a small range of different block types, redundant blocks of code which did not function at all, and contained large numbers of ‘Sprites’ (‘Sprite’ refers to a character, or object in a Scratch program).

Students generally did not understand why the blocks they were using caused their programs to behave the way they did and were more interested in making animations. A common example of this type of program would be one containing ten different sprites where one of these rapidly spun in a circle, a very exciting thing for a student to watch and have created, but not something particularly instructive. When left to continue exploring without guidance students commonly did not progress to using different blocks or trying new things. Often once they had put together a block of code they were happy with they would reuse this frequently and would not build on it further. While this did function as a creative outlet for students, it was not building any programming or CT skills. This reflected what Aivaloglou and Hermans (2016) would find in their future work analysing large data sets of Scratch projects from the Scratch repository [2].

To combat this tendency an emphasis was put on teaching programming concepts *with* Scratch, rather than *teaching Scratch*. Students were explicitly told that they were going to be learning programming and programming *with* Scratch, and that while there was nothing wrong with creating simple animations they were not going to focus on that during class time. They were welcome to do this at any other time though and were encouraged to be creative and experiment outside of the programming focused classes. A small number of students had previously used Scratch, and some had used this extensively at home. Though these students were more confident with using the Scratch interface they generally had no more prior knowledge of programming practices than other students and were still prone to ‘Sprite-fest’ style programming.

Angela based the programming lessons on resources I wrote for her, and activities I guided her through when teaching her. These resources and one-on-one tutorials covered the programming concepts of inputs, outputs, sequences, variables, selection, and iteration. As Angela came to better understand these concepts she also began searching for Scratch teaching resources online and adapted many activities to create new lessons. It is worth noting here that within two years of beginning to learn programming using Scratch Angela felt she had truly grasped these programming concepts and, from my observations, this was entirely accurate. She was confident in saying that she understood how to code.

The programming activities used covered a range of subject areas. Students worked on programs that performed basic mathematics (such as multiplication and division), a number guessing game, and drew geometric shapes. They did a project on the 2014 Winter Olympics, incorporating literacy and geography, in which a character moved between countries, greeted the user in the national language, and reported a fact about the country. Towards the end of term 3, and during term 4, students were given a final project to create an interactive quiz about a native New Zealand bird. Angela chose this as a project because it exercised students programming and algorithmic thinking skills, and brought together several different learning areas. Students had to: research a native NZ bird and collect facts about it, create and draw a sprite for their chosen bird in Scratch, practice their literacy skills when writing the quiz, and carefully consider the user of their program when they built it to ensure it was usable. Implementing this fully required each of the programming concepts covered in class: inputs, outputs, sequences, variables, selection, and iteration. Not all students successfully completed this project, but all were able to implement sections of it with teacher assistance. These projects were displayed to parents and whānau at a school open day.

7.5 Assessment Process

Five different methods were used to collect information about the effects of the study on students and Angela. This information was collected to address the research questions listed in section 7.2

Teacher’s professional opinion: Throughout the programme, I remained in close contact with Angela via email and frequent meetings. During this time we discussed her observations of the classes, students reactions to and engagement with the material, her own experiences with the material, and her general opinions about including these topics in her teaching program. I recorded these through note-taking and email records between us. This information was collected to try to answer research questions 1 through 7.

Bebras challenge: Students took the 2014 Bebras challenge after they had been studying the computing topics. This was done to gather information on students CT and problem solving skills, to address research questions 4 and 7. I originally intended to compare the results of this to results for the same year group in NZ, but I was unable to do this, as no other students of the same age group took the challenge that year.

Binary numbers quiz: This was developed by Angela and me. It was designed to test students' basic understanding of the concepts covered by the Data Representation CS Unplugged activities, to address research questions 3 and 7.

Scratch/programming quiz: This was developed by Angela and me. It was designed to test students' basic understanding of programming using Scratch, to address research questions 3 and 7.

Survey of student attitudes: At the end of the school year, after all classes had taken place, students were asked to complete an attitude survey. This asked if they enjoyed the programming and data representation classes, their perceived ability in these subjects, and opinions on working in programming and Computer Science in the future, to address research questions 5, 6, and 7.

Google Forms⁸ was used to administer the two quizzes and attitude survey. This was a choice made by Angela. She chose to do this as the school used Google Classroom⁹, she used Google Forms and Google Classroom frequently for all topics she taught, and so her students were familiar with using it.

Due to the logistical issues which tend to occur when dealing with large numbers of participants in an unpredictable environment (such as a school), not all students completed the Bebras challenge, each quiz, and the attitude survey.

As previously mentioned, Angela was also interviewed in 2016, as she had continued teaching these topics, and had created many new lessons and projects based on her increased knowledge. Although this interview was not a part of the pilot study, it will be discussed in the results section of this chapter, as it provides many insights into Angela's initial experiences and the longer-term impact of this study.

7.5.1 *In-class quizzes*

To assess how much students had learnt from the programming and CS Unplugged lessons, they completed two short quizzes that covered some basic programming and data representation concepts. Each of the quizzes administered was multi-choice and written by Angela.

⁸ <https://www.google.com/forms/about/>

⁹ <https://edu.google.com/k-12-solutions/classroom/>

She discussed the questions she had chosen with me, and I checked that the wording and answers were technically correct.

Some examples of questions from the data representation quiz are: converting numbers between binary and base-10, taking a set of binary numbers and converting these to letters in order to read a message. It also included some of the more abstract concepts they had discussed in class, such as the largest and smallest numbers you can represent, and which whole numbers they could make, with a given number of bits. Some examples of questions from the programming quiz are: what a variable is, the concepts of inputs and outputs to programs and examples of these, and interpreting a short Scratch program to determine its result. The full sets of questions and answers for these quizzes are listed in Appendix B.

7.5.2 *Attitude survey*

The questions in this survey were reviewed and approved by the University of Canterbury Educational Research Human Ethics Committee.¹⁰ At the end of the year students were asked to respond to the following statements on a 5-point Likert scale, from Strongly Disagree to Strongly Agree:

Q1. “I enjoyed learning and using Scratch in class”

Q2. “I enjoyed learning about how data is represented with binary”

Q3. “I want to continue learning about Programming and Computer Science”

Q4. “I will consider working in Programming or Computer Science as a job”

Q5. “Before this year, I would have considered working in Programming or Computer Science as a job”

Q6. “I am good at programming with Scratch”

Q7. “I am good at understanding and using binary numbers”

Students were also asked to enter their gender and select the ethnic groups that they belonged to. These questions were asked at the end of the quizzes once students had attempted

¹⁰<https://www.canterbury.ac.nz/study/ethics/educational-research-human-ethics-committee/>

all questions, to lessen the risk of stereotype threat impacting students answers [58]. When entering their gender the options given were ‘Female’, ‘Male’, ‘Other’, and ‘Prefer not to answer’. When choosing which ethnicity options to include in the survey we selected the options used by NZ Stats¹¹ when conducting the NZ census, with the addition of ‘(Pākehā)’ to the ‘New Zealand European’ option. The options were ‘New Zealand European (Pākehā)’, ‘Māori’, ‘Samoan’, ‘Cook Island Māori’, ‘Tongan’, ‘Niuean’, ‘Chinese’, ‘Indian’, ‘Prefer not to answer’, and an ‘Other’ option was given where students could type in other ethnic groups they belonged to. Lastly, students provided their age.

Before analysis, entries where students had selected ‘Other’ for their ethnicity and entered one, were reviewed and converted to the matching category from the NZ census options. For example, the entry ‘French’ was replaced with ‘European’. All entries were successfully matched.

Potential biases

Although students were asked to fill in their gender at the end of the survey, they were told they would be asked about this at some point as Angela explained to them that there would be an ‘other’ option in the gender question and why this was there. This was to discourage students from simply choosing ‘other’ because they found it amusing, a risk with students of this age. This meant that although students did not input their gender before they began the survey, they were still made aware of it, which may have influenced their answers due to stereotype threat.

7.6 Results

In total, data was collected from 599 students: 310 year 7s and 289 year 8s. However, there was not a complete data set for each student. A complete set contained: scores for both in-class quizzes, a score for the Bebras challenge, the students gender, the ethnic groups they belonged to, their answers to the attitude survey, and their age. The majority of students completed a subset of these. The year level for all students was collected. The amounts of each type of data collected are shown in tables 7.1, 7.2, and 7.3.

In analysing the data from the Bebras challenge, the two in-class quizzes, and the attitude survey I aimed to address the following five of the questions listed in section 7.2:

¹¹ <https://www.stats.govt.nz/>

Data type	Valid Data	Missing Data
Gender	554	45
Ethnicity	462	137
Survey Response	473	126
Bebras Score	226	373
Binary Score	366	233
Programming Score	375	224

Table 7.1: Amount of data collected for each measure

	Female	Male	Other	Unknown
Year 7	140	141	1	28
Year 8	131	138	3	17
Total	271	279	4	45

Table 7.2: Numbers of students, by gender, in each year group

	Māori/Pacifica	Non Māori/Pacifica	Unknown
Year 7	48	169	93
Year 8	52	192	45
Total	100	361	138

Table 7.3: Numbers of Māori/Pacifica students per year group

3. Whether students in a typical classroom could be taught CS and programming concepts, through their teachers use of CS Unplugged resources, and teaching of block-based programming.
4. Whether students developed CT skills, and general problem solving skills, through this process.
5. How students reacted to, and engaged with these lessons.
6. The degree to which students enjoyed the CS Unplugged and programming lessons, how confident they were in their abilities in programming and the CS Unplugged topics, and if they wanted to continue learning about them.
7. Whether these observations and results varied for students of different genders, or different ethnicities.

7.6.1 *Bebras*

The data obtained from Bebras included students' names, gender,¹² and their final score. Students marks for each individual question were not available. For a description of the Bebras marking rubric see Chapter 6, section 6.4.1. The results of the Bebras challenge followed a normal distribution, and are shown in Figure 7.1

When students' scores were categorised according to their year level, gender, and whether they identified as Māori/Pacifica, the scores were also normally distributed. These distributions are described in tables A.8, A.9, and A.10, in Appendix A.3.1.

An Independent samples T-test was performed to test for differences in students' scores for the groups: Female vs male students, for year 7 vs year 8 students, and for Māori/Pacifica students vs non-Māori/Pacifica students. The results of these showed no statistically significant differences between female and male students ($p > 0.05$), between year 7 and year 8 students ($p > 0.05$), or between Māori/Pacifica students and non-Māori/Pacifica students ($p > 0.05$).

¹² The options provided by Bebras were 'male' and 'female'

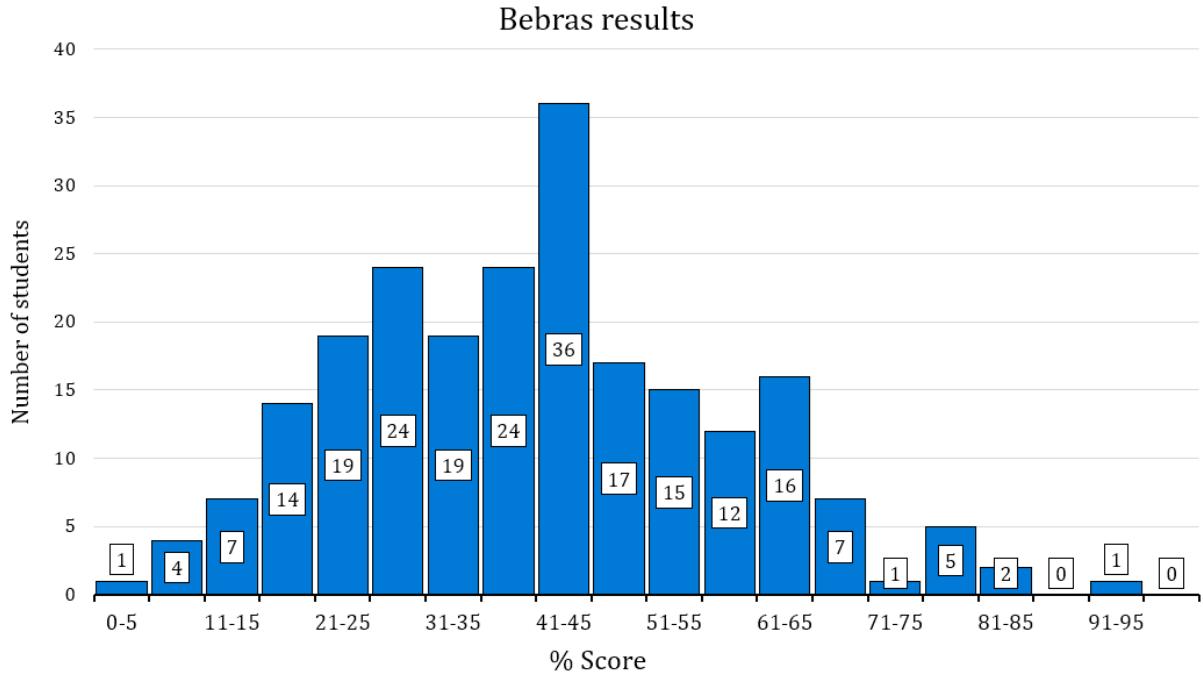


Figure 7.1: Distribution of students' scores in the Bebras challenge.

7.6.2 Data representation Quiz

The distribution of student scores is shown in figure 7.2. Students' scores divided by their year level, gender, and whether they identified as Māori/Pacifica, were also normally distributed. These distributions are described in tables A.11, A.12, and A.13, in Appendix A.3.1.

An independent samples T-test was performed to test for differences in students' scores for the groups: Female vs male students, for year 7 vs year 8 students, and for Māori/Pacifica students vs non-Māori/Pacifica students. The results of these showed no statistically significant differences between female and male students ($p > 0.05$), and between Māori/Pacifica students and non-Māori/Pacifica students ($p > 0.05$).

However, there was a statistically significant difference between the scores of students in year 7 and those in year 8 ($p < 0.05$). The effect size of this difference was 0.57, indicating the year level had a medium effect on students' scores. Surprisingly students in year 7 scored higher than students in year 8. It would be expected that older students, who have spent more time in school, would have higher scores than younger students. It is also unexpected as this difference between year levels was not present within either of the other tests. One

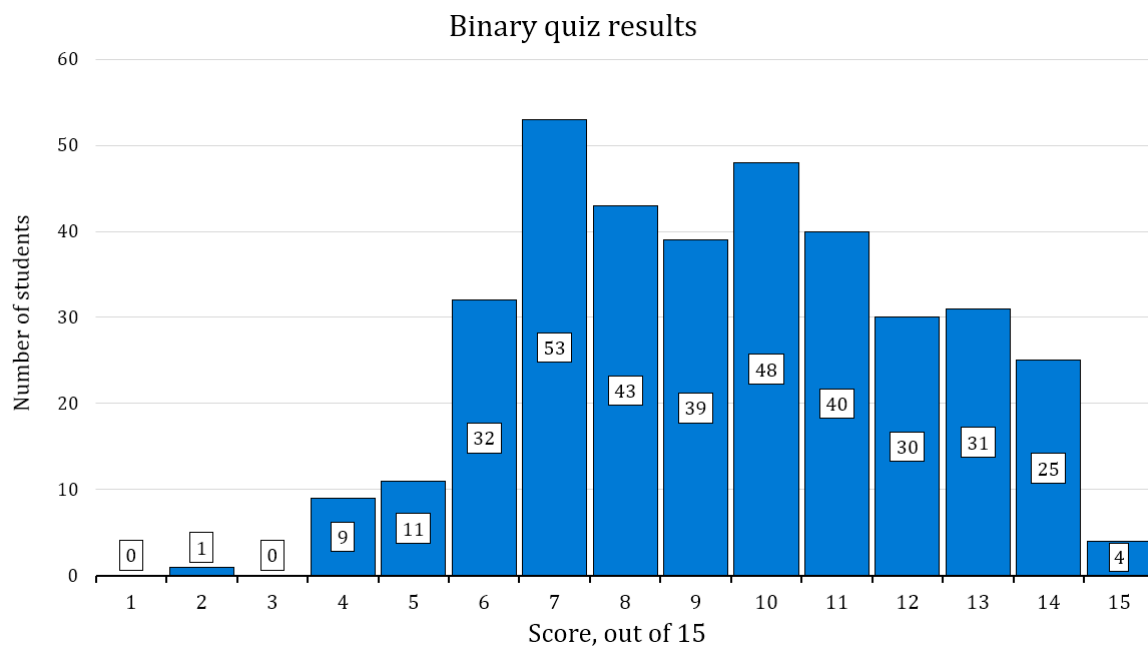


Figure 7.2: Distribution of students' scores in the Binary quiz.

possible explanation is the timing of school camps, as groups of students from different year levels were away intermittently throughout the year, and more year 8 students than year 7's may have been away and missed specific binary classes. Another is simply the timing of lessons, as the time between doing these lessons, and doing the test, may have differed between year groups, with year 8's learning this content earlier in the year than year 7's.

7.6.3 Programming Quiz

Before analysing the programming quiz results several changes were made to it. One question was removed as it was unrelated to the lesson content (Angela taught the content it related to separately and wanted an opportunity to assess it for report writing purposes). Questions one and four were remarked as it was later deemed the wording in these was ambiguous and so two of the possible answers could be considered correct. The distribution of student scores is shown in figure 7.3

Students' scores divided by their year level, gender, and whether they identified as Māori/Pacifica, were also normally distributed. These distributions are described in tables A.14, A.15, and A.16, in Appendix A.3.1.

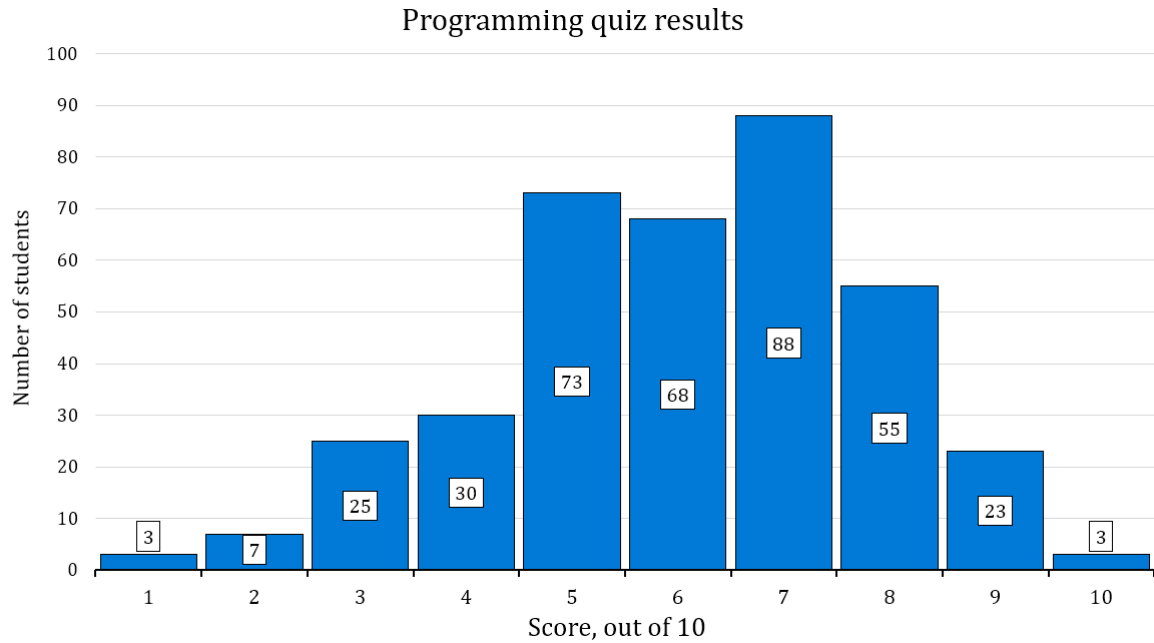


Figure 7.3: Distribution of students' scores in the Programming quiz.

An Independent samples T-test was performed to test for differences in students' scores for the groups: Female vs male students, for year 7 vs year 8 students, and for Māori/Pacifica students vs non-Māori/Pacifica students. The results of these showed no statistically significant differences between female and male students ($p > 0.05$), between year 7 and year 8 students ($p > 0.05$), or between Māori/Pacifica students and non-Māori/Pacifica students ($p > 0.05$).

7.6.4 Quiz comparisons

There were statistically significant correlations between students' scores in each of the quizzes and in the Bebras challenge. These correlations are shown in table 7.4. However, the correlation between the Bebras scores and programming quiz scores did not hold across all sub groups of students.

This correlation was only marginally significant for year 8 students ($p = 0.06$), and for male students ($p = 0.07$). For Māori/Pacifica students the significance of a correlation between these scores was far from the 0.05 level ($p > 0.5$). As there was no significant difference between Māori/Pacifica students, and non-Māori/Pacifica students' scores in Bebras and

		Bebras score	Binary total
Prog total	Pearson Correlation	0.241	0.424
	Sig. (2-tailed)	0.007	< 0.001
	N	125	329
Binary total	Pearson Correlation	0.515	1 (N/A)
	Sig. (2-tailed)	< 0.001	
	N	131	

Table 7.4: Pearson correlation between scores in each of the quizzes.

programming, this lack of correlation is unexpected. It is possibly due to the small number of Māori/Pacifica students who completed both the Bebras and programming quizzes (N=18).

7.6.5 Attitude Survey

For analysing the results of the Attitude Survey, answers were coded with the values 1 - 5, with 1 representing ‘Strongly disagree’ and 5 representing ‘Strongly agree’. For the sake of brevity, in the following figures and discussions the questions will be referred to by either their numbers only, or in the following way: **Q1)** Programming enjoyment, **Q2)** Data-rep enjoyment, **Q3)** Continue learning, **Q4)** Consider a career, **Q5)** Prior career interest, **Q6)** Programming confidence, **Q7)** Data-rep confidence. Figure 7.4 shows the answers students gave to the survey.

Students generally had neutral to positive feelings about the programming and data representation lessons, and their own levels of competence in these topics. Only 17.1% of students stated they did not want to continue learning about these subjects.

While overall attitudes towards careers in programming and CS were generally negative, the difference between students answers to **Q5)** and **Q4)** showed a change in students attitudes towards these careers. As CS is one of many possible career paths it would be unexpected, and undesirable, to have absolutely everyone pursue this path. The proportion of students who were not interested in this career path is therefore not particularly surprising or discouraging. Almost one quarter of students (24.3%) reported that they would consider a career in this area. The number of students who disagreed, or strongly disagreed, with the possibility of pursuing this career path decreased from 56.4% to 46.3% after learning about these subjects.

A Spearman’s correlation test was used to determine the relationships between students answers to each survey question. Unsurprisingly, there were significant positive correlations

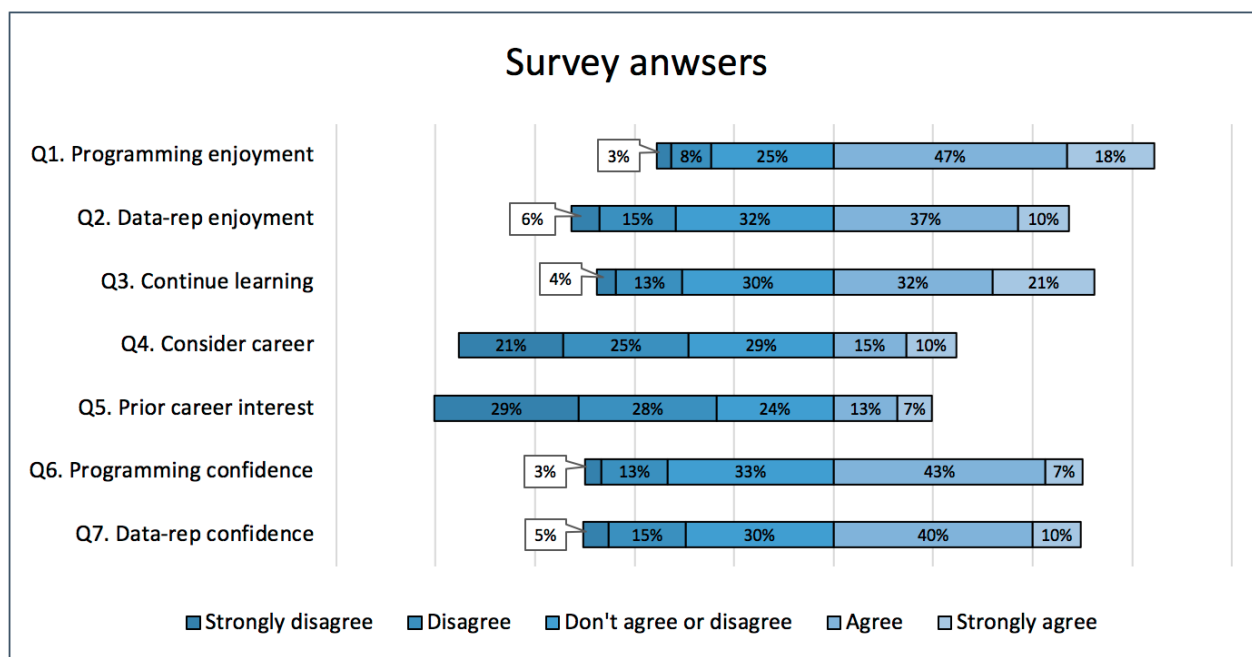


Figure 7.4: Students' answers to the attitude survey administered at the end of the study

between students' answers. These ranged from weak (0.29) to strong (0.66), and are shown in table 7.5. These correlations held for different year levels and genders, as shown in tables A.18 and A.17, in Appendix A. These correlations held for Māori and Pacifica students, except for one case: **Q1)** 'Programming enjoyment' and **Q5)** 'Prior career interest', this correlation was only marginally significant ($p < 0.10$). These correlations are shown in table A.19, in Appendix A.

Students who were confident in their skills with programming were more likely to enjoy programming, and likewise for students who were confident in their understanding of binary, and enjoyment of it. The more confident students were in one subject, the more likely they were to be confident in the other, and enjoy each of them. Students who were more confident and enjoyed the topics were more likely to want to continue learning CS and programming, and more likely to consider working in it.

Differences between year groups, genders, and ethnicities

A Mann-Whitney test was used to investigate if there were significant differences between the survey answers for different groups of students. This indicated that students' answers to the survey were not correlated with whether or not they were Māori/Pacifica ($p > 0.05$

Questions	Q2	Q3	Q4	Q5	Q6	Q7
Q1) Programming enjoyment	0.51	0.51	0.41	0.29	0.54	0.34
Q2) Data-rep enjoyment		0.48	0.45	0.39	0.41	0.55
Q3) Continue learning	0.48		0.60	0.42	0.40	0.40
Q4) Consider a career	0.45	0.60		0.66	0.42	0.38
Q5) Prior career interest	0.39	0.42	0.66		0.31	0.29
Q6) Programming confidence	0.41	0.40	0.42	0.31		0.44
Q7) Data-rep confidence	0.55	0.40	0.38	0.29	0.44	

All correlations significant at the $p < 0.001$ level.

Table 7.5: Correlations between all students' answers to the attitude survey questions.

for all questions). When comparing students in year 7 and year 8 the results of this test showed no evidence of differences in questions 2, 4, 5, 6, and 7 ($p > 0.05$). However, there was a significant difference in students' answers for **Q1)** 'Programming enjoyment' ($p < 0.05$), and **Q3)** 'Continue learning' ($p < 0.05$). In both of these cases year 7 students were more likely to answer positively. This indicates that they were more likely to enjoy programming with Scratch, and more likely to want to continue learning about these topics. A possible explanation for this is the older students may have been more likely to view Scratch as something 'childish' and not suitable for their age group. While the effect of students year level on these questions was statistically significant, they were small. The effect sizes were 0.255 for **Q1)**, and 0.187 for **Q3)**.

When comparing students of different genders, the results of the Mann-Whitney test showed significant differences between female and male students' answers to all survey questions (in all cases $p < 0.01$). The distribution of female and male students' responses is shown in figure 7.5. The effect sizes of these differences ranged from 0.27 - 0.58. All but one of these effect sizes (0.27, a small effect, for **Q7)**) indicate a medium effect of student's gender on their answers. The full table of effect sizes can be found in table A.20 in Appendix A.

Female students were more likely to answer negatively to all survey questions. This indicates that they had lower confidence in their abilities, enjoyed the topics less, were less likely to want to learn more, and less likely to consider working in CS and programming,

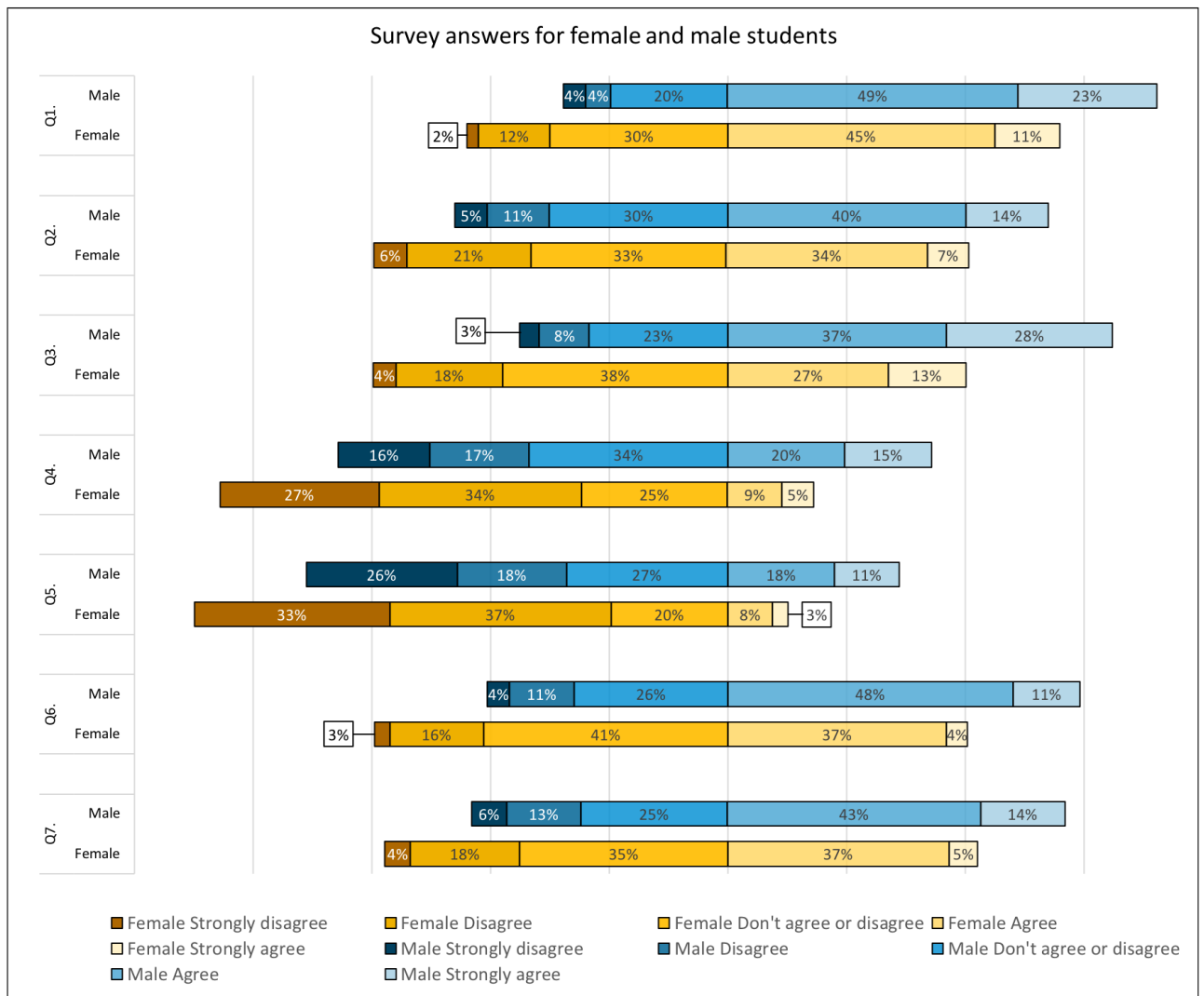


Figure 7.5: Students' answers to the survey questions, divided by gender.

both before and after the course. This trend of girls' and boys' differing attitudes towards these subjects has been well documented across a range of ages, countries, and education systems. It appears the program of work conducted in this school was no exception to these trends.

I further investigated these differences between genders and age groups. I compared female to male students within year 7, and within year 8. I also compared year 7 and 8 female students to each other, and year 7 and 8 male students to each other. These comparisons were again done using the Mann-Whitney test. The full results of these tests can be found in tables A.21-A.24 in Appendix A. This revealed several statistically significant differences between year groups, which differed from the results when comparing all female and all male students, shown in A.20.

When comparing female year 7 students to female year 8 students, significant differences were found in their answers to **Q1)** 'Programming enjoyment', and **Q3)** 'Continue learning'. Year 7 female students were more likely to answer positively to each of these questions. The effect sizes of these differences were 0.31 and 0.3 respectively, indicating a small effect. In contrast to this, there were no statistically significant differences between male year 7 and year 8 students answers to any survey questions.

When comparing female year 7's with male year 7's, and comparing female year 8's with male year 8's, there were statistically significant differences between female and male students' answers for questions 1, 3, 4, and 5. The effect sizes of these correlations were calculated for each group, in order to compare the differences in the impact of gender in the two different age groups. The full result of this can be found in Appendix A.3.2, tables A.21 and A.22. For both age groups, gender had a small and very similar impact (0.34 and 0.37) on **Q1)** 'Programming enjoyment'. For **Q3)** 'Continue learning', and **Q4)** 'Consider career', gender had a medium effect on students' answers in both age groups. However, in both cases the effect size for year 8's was larger (0.42 vs 0.59 for **Q3)**, 0.49 vs 0.67 for **Q4)**. There was also a difference of effect size for **Q5)** 'Previously considered career', between the age groups, with Gender having a small impact on **Q5)** for year 7's (0.38), and a medium impact for year 8's (0.65).

Opinion changes on working in CS and/or programming

Individual students' answers to questions 4 and 5 of the survey were compared, to establish if their opinion on working in programming or CS had changed over the course of the program.

Student group	Negative change	No change	Positive change
All	12.1%	55.9%	32.0%
Year 7	10.7%	61.6%	27.2%
Year 8	13.3%	50.8%	35.9%
Female	11.9%	59.0%	29.1%
Male	11.7%	52.7%	35.6%
Māori/Pacifica	14.0%	54.0%	32.0%
Other ethnicity	11.4%	56.9%	31.7%

Table 7.6: Opinion changes on working in CS and/or programming

The results of this for all students, and relevant subgroups, are shown in table 7.6. There were no significant correlations between opinion changes and any of: gender, year group, scores in any of the assessments, and Māori/Pacifica students ($p > 0.05$ for all).

7.6.6 Comparing Survey Responses with Assessment Results

Students' results in each assessment were compared to their answers to the survey, to investigate the impact of their achievement in these areas on their opinions of CS and programming. The results of this are shown in Appendix A, tables A.25 and A.26. The only questions that had a statistically significant correlation with students' quiz results were those on enjoyment and confidence, but in each case these were relatively weak (ranging from 0.13- 0.19). No significant correlation was found between the survey answers and students' Bebras scores ($p > 0.05$).

Confidence and achievement

The relationships between student's confidence in their abilities and their scores in the assessments were also investigated with students divided by gender, and by ethnicity. The full results of this are shown in tables A.27 and A.28 in Appendix A.

In all cases there was no significant correlation between students confidence and their score in the Bebras challenge. For boys there was a weak-medium positive correlation between their level of confidence and their scores in the Binary and Programming quizzes. However, for girls there was no significant correlation between their confidence and their scores.

Similarly there was no significant correlation between achievement and confidence for Māori/Pacifica students, but there was a correlation for non-Māori/Pacifica students. In

both cases this indicates that these groups generally had an inaccurate perception of their abilities. These results are interesting as there was no correlation between gender and scores, and no correlation between whether a student was Māori/Pacifica and with their scores. There was also no significant correlation between whether or not a student was Māori/Pacifica, and with their confidence.

7.7 Discussion

Before the course began, Angela had felt apprehensive and nervous about the classes but found that after the first few weeks of lessons she had become more confident. Having one teacher learn all the content and present it to students offered several benefits. It allowed me to provide plenty of one-on-one professional development, react quickly to observations Angela and I made about the material, and adapt the support I provided. It also meant that Angela was teaching the same lesson to multiple classes each week, which quickly built her confidence in how to deliver the content. It also allowed her to gain a deeper understanding of the concepts she was teaching, and she stated that she gained a much better ‘big picture idea’ of how the full course of lessons fitted together.

Throughout the year as her knowledge grew, she began having her own insights into the topic, how it could be taught, and made connections to other subjects that my research group and I had not previously thought of. This illustrates the significance of the skills experienced teachers bring with them to new subject areas. There is a stereotype that computing is an area ‘just for young people’. However, through this pilot study (and the following studies) it was clear that the age of a teacher is not an automatic disadvantage, and can be an advantage because of the wealth of knowledge and experience that they bring to the subject.

Teaching this did, of course, require a huge amount of new learning for Angela. She told me that the year had been a stressful and particularly demanding one, even though she had a “massive amount of support” from her school, her principal, and me. This had the potential to cause isolation from other staff, but Angela did not struggle greatly with this as she had been teaching at the same school, and teaching a subject only she did, for many years. In what would become a recurring theme throughout my research, she said *time* was the biggest challenge she faced. It was not simply an issue of finding time to learn the content, but also time to learn how to present it. It also presented challenges for assessment as, with over 600 students in total, individual evaluation of students was not possible without some form of

standardised testing.

Despite this, Angela said it had still been a positive experience for her and her students, and she was happy that she took on the challenge. At the end of 2016, after she had been teaching this for three years I asked her “If you could go back to the beginning before you started this, would you still do it?”. Her answer was a definite yes, and she felt she had been “lucky to be able to do this”.

7.7.1 Aims of the programme

At the beginning of the year, Angela and I had agreed upon the aims (listed in section 7.2) of running this programme and were pleased to find at the end of the year that we had achieved all of these.

Students had enjoyed, and been engaged with both the CS and programming content throughout the year, and Angela believed that each student had been able to achieve something, no matter how small or large, in her classes. The students who had excelled with the material had been directed to extension resources, and Angela felt she had been able to encourage and support them, without having to take extra time away from the students who required more help. Students had been collaborative and shared their learning with others when they needed assistance, particularly with the programming material, and the extension students were often involved in teaching and supporting others. Many students had also gone home and shared their learning with their families, teaching their parents and caregivers binary and performing the parity magic trick. Students who worked on programming at home or at lunchtime would ask their friends to play the games they had created, and try to improve each other’s games. Angela was able to use the assessment results, and her general observations in class, to give a general indication of students achievement throughout the year in their end-of-year reports.

7.7.2 Gender differences

High engagement and enthusiasm from students was observed from very early in the programme, and this also came out in the attitude survey. When initially asked about whether all students or only some groups of students displayed this interest, Angela stated she did not observe any difference across different ethnic groups but did see some difference between girls and boys. It was observed that in the early stages of the course several girls seemed to take longer than the boys to display enthusiasm for, and understanding of the concepts.

However, as time went on Angela observed another difference between this specific group of girls, who took longer to complete some activities, and the rest of the students. They seemed to require more encouragement when confronted with this new content, were anxious about starting the activities in case they did something wrong, and were less willing than other students (particularly the boys) to take risks. They were also less persistent. Their struggles weren't because they found the content more difficult than other students did. The cause instead appeared to be their fear of failure. This 'fear of failure', and tendency to avoid taking risks has been documented as more common for both girls and women, in a variety of areas, and specifically in computing [123]. It is also strongly associated with perfectionism, and observations similar to this were made by many teachers throughout the following studies.

An example of this behaviour is how these girls responded when given the 'Run-length encoding' activity to complete. This activity involved using an algorithm to interpret a set of numbers, which told students which particular squares on a grid they needed to colour in. When this was done correctly the squares formed an image. While the majority of students grasped the concept of the activity and were able to complete all of the problems on the activity sheet, some of the girls seemed to struggle and completed only one or two of the images. This was not a reflection on their understanding of the task though. It was observed that they made less progress than the other students as they became hyper-focused on filling in each square completely, so that no white paper would show through, before they moved on to the next square. This behaviour persisted even after the teacher and I told them multiple times that this was unnecessary, not the point of the exercise, and a waste of their class time. From my perspective, they had less self-confidence in their performance at the end of the class than the rest of their classmates. The significant differences between girls and boys answers to the attitude survey, combined with the lack of difference in their test scores, back up these observations. It was not ability levels that set the girls and boys apart, but social and personal influences.

However, Angela's observations about gender differences changed over the following years. As highlighted in Chapter 2, changing these differing attitudes and experiences of the subject between students of different genders is one of the goals of introducing these topics to primary school. From the interview I conducted with Angela in 2016, it seems that this effect may have been achieved. In the interview I tasked about these observations she had made, and whether had persisted or changed. She answered that when it came to interest and aptitude in the subject.

“It was probably 50/50 really. If I think of the ‘extension’ programmers, there would be as many girls as boys... I don’t see really any difference between the boys and the girls”.

This was a very interesting and encouraging result.

7.7.3 Teaching methods

Angela’s experiences provided insights into the ways these materials can be taught and used, both by themselves and alongside other subjects. She felt that for the course to be more successful the lessons needed to be reinforced more. She suggested doing multiple small projects as students progress through the material, exercising the same skills frequently with slightly different material e.g. using different numbers, words, or mathematical functions.

7.7.4 Data Representation

Angela found that students were “hugely engaged” by the data representation material, and their excitement was particularly apparent when students began asking “What are we doing next week?”, at the end of every class. She felt that the students had developed an understanding of the binary number system, how binary digits can be used to represent different types of information, and the more abstract concept that computers represent all data in digital form.

She experimented with using more than ones and zeroes to represent binary numbers, and had students decode ‘secret messages’ written using different shapes (such as clubs and spades), Māori words, and koru patterns. This reinforced the learning and helped students understand that the concept was not just about numbers or zeroes and ones, but it was about using two states (or objects) to represent information.

Students were also very interested in the error detection and correction material. Initially, Angela felt this was just because of the activity was presented as an exciting magic trick. However, as time went on most students were able to connect this concept to parts of their everyday lives. They understood some of the importance of this process to how digital devices in the world around them worked.

7.7.5 Programming

With specific regard to Scratch, Angela wanted to incorporate more play and directed exploration into the initial lessons when students were first introduced to the language. She felt this would help all students build some confidence in *using* Scratch, and become more comfortable with the interface and program itself. This would be helpful for students to have before they moved into the programming concepts themselves. In almost every class there were a small number of students (typically 1-3) who had previously used Scratch outside of school. These students were, unsurprisingly, much more confident than others and more enthusiastic than most. Although many of these students had used Scratch as more of an animation tool and did not have more specific programming knowledge than others, their familiarity with the language was beneficial. The majority of these students were boys. This illustrates again the impact and importance of prior experience with these subjects, and particularly with programming environments. In some cases, these students became ‘classroom helpers’ and assisted other students. This behaviour will be discussed in Chapter 8, as it came up frequently in the interviews conducted with other teachers, as well as with Angela.

Angela found that incorporating maths and geometry with programming was highly successful. For example, when students were working on drawing shapes in Scratch by using the pen mode and moving a sprite around the screen, they needed to use specific angles when drawing regular polygons. They began asking more about geometry and became more engaged in their maths lessons on the topic. Their motivation and engagement in programming transferred to the maths curriculum because they wanted to understand how to do different things in Scratch. Because of this she progressively used Scratch more and more when teaching maths in the following two years.

Towards the end of the year, Angela gave students a Scratch programming project to complete. They were tasked with creating a quiz about a native New Zealand bird. Angela decided on this project as creating a quiz would require students to understand and apply a wide range of core programming skills: sequential thinking; using input and output commands; variables; conditional statements; and loops. It also required students to think about the user experience and was a cross-curricular activity. It incorporated: art and design, with students creating their own sprites in Scratch; literacy, with students writing their questions and answers; and inquiry learning, as students researched NZ’s endangered native birds. While this would have been a good way to assess students learning, it was not possible to do

this for individuals due to the vast number of students taught, and the time it would take to assess each project.

The amount of success students had, and the progress they made through the task, varied greatly. Some students continued to add new features to their projects long after they had satisfied the requirements, and others struggled to get more than a few questions working in their quiz. However, Angela said that every single student was able to achieve and produce something, and she felt very happy with their results. Many of these projects were then presented by the students at a school open day, for their whānau and visitors to see.

7.8 Conclusion

Working with Angela provided a solid foundation to build my future research on. Together we had found that despite being brand new to these topics, she was able to effectively integrate these into her classroom in a way that was beneficial for students learning. Our goals for the programme had been achieved, with students being highly engaged, and able to learn the content covered. The main outcomes of this pilot study were:

- My understanding of how a teacher could be supported to integrate this material had increased, and we found that the materials used could be successful in an intermediate school environment.
- As expected based on prior research the CS Unplugged activities were suitable for this age level. Students were highly engaged by, and successful with the content.
- From our assessments and Angela's observations, it appeared there was no significant difference between the achievement of students of different genders or ethnicities.
- Thorough measurement of students' learning was, as expected, difficult, as a scalable strategy for doing this had not yet been found.
- There were significant differences in girls and boys attitudes towards programming and CS, with girls reporting lower enjoyment and interest in the area.

After the success of this programme, I wished to investigate whether this could be emulated in different school environments, with a range of teachers and age groups of students. How to support larger numbers of teachers was also a subject of interest, as implementing

this subject in the national primary curriculum would create a need for large scale training and resourcing for teachers. When examining different age groups of students I wished to investigate whether the material could be made suitable, engaging, and would be beneficial for all ages; and if the differences between girls and boys attitudes to CS and programming changed across year levels. These topics are explored in the next chapter, through the work undertaken with a larger group of NZ primary teachers.

Chapter VIII

Open Teachers Study

After working with Angela and seeing the success she had teaching data representation, programming, and algorithms, I investigated whether these observations held across different environments. To do this I began working with a larger group of teachers in 2015. I intended to conduct a one year study with this group. However, during this year the demand for professional development (PD) in this area for primary school teachers rapidly increased. More primary schools were recognising the need for this content and learning that it would be added to the curriculum soon. This resulted in the UC Computer Science Education Research Group rapidly expanding our teacher support and PD efforts. Through this, more teachers heard about my research and asked to participate in it. Because of this, I chose to extend this study to the end of 2016, and by the time it ended I had worked with 40 teachers at 25 different schools, and students ranging from 3 to 13 years old. This chapter covers the work done over these two years.

In this chapter I will cover how this study was conducted, the research questions I aimed to answer, how I collected information from teachers throughout their time teaching CS and programming, and how this qualitative data was analysed. The motivations and experiences of the participating teachers, the specific lessons which were taught, and the Computer Science concepts these covered are described. I then discuss each of the key themes which were identified through teachers feedback and interviews: Computational Thinking, The Key Competencies, problem solving, Cross-curricular Learning and Learning Transfer, and Impacts on Students. I conclude by addressing my original research questions for this study and identifying the main area of focus for my final study.

8.1 Overview

This project aimed to further gather information on the impacts of integrating Computer Science, Programming, and Computational Thinking material into primary school class-

rooms, on students and teachers. Participants were trained in teaching CS and CT through CS Unplugged, programming Unplugged, programming using block-based languages (e.g. Scratch and Scratch Junior), and programming floor robots (e.g. Bee-Bots). This was done through the ‘Computer Science for Primary Schools’ (CS4PS) workshops and in some cases one-on-one support from myself and another member of the CSERG, Wendy.¹ Wendy is a primary school teacher, who participated in this study by submitting feedback on lessons she conducted. During the time she participated, she also joined the CSERG and became a core contributor to resource development and teacher training. Along with conducting lessons and submitting feedback on these, she worked to support five of the other participants in this study and served as a mentor to many of the teachers she worked with.

A variety of recruitment methods were used, primarily advertising at CS4PS workshops and other professional development events in Canterbury. Information about the study and invitations to participate were posted on the Facebook Group ‘NZ Teachers (primary)’. Primary schools and teachers we had existing relationships with were also invited to join, and several teachers contacted us directly after hearing about the study from their colleagues.

Teachers who participated were asked to provide feedback on the lessons they used in class via an online form. As the study progressed, I decided to extend my data collection and gather more information on the teachers involved. I asked participants to fill out a survey about their motivations for teaching this content, participating in the programme, their prior experience with this material, and their personal feelings about the programme. At the end of 2016, I conducted interviews with 18 of the participants. These each ranged from approximately 30 to 75 minutes in length

Recruitment began in mid-2015 and continued through to the end of 2016. Teachers were able to join or withdraw from the study at any time. Because of this, the amount of time people participated varied for each individual. Several participants continued teaching and providing feedback throughout the entire project, but the majority participated in either 2015 or 2016, not in both years. Angela (the teacher I worked with in the 2014 Pilot study) also participated in this study.

While carrying out this study, I was also taking part in national discussions and working groups focused on changing the NZ Digital Technologies curriculum. Consequently, I was aware of the planned extension of DT into the primary school curriculum and was able to contribute to the draft versions of this. These plans and the contents of the draft curriculum

¹ Name has been changed

went through many iterations and changes over this time. These changes, and my role in the curriculum drafting, continually influenced the work I did with teachers during this study. In turn, this work with teachers also influenced my contributions to the curriculum changes. This meant the specific content I taught participants, and the resources I asked them to trial, shifted over time. As many of the teachers involved were also aware of the planned changes in the curriculum, this influenced their own goals and motivations for teaching this material.

8.1.1 Research Questions

The research questions I aimed to answer in this study were:

1. Can Computer Science and Computational Thinking skills be taught in a typical primary school environment? (RQ1)
2. How suitable are the teaching resources that were chosen for use? (supports RQ 1.2 and 1.3)
3. Can learning Computer Science and programming develop, or improve, Computational Thinking skills for the majority of students? (RQ2)
4. Can learning Computer Science and programming, and developing Computational Thinking skills, have an impact on problem solving skills or general learning, for the majority of students? (RQ3)
5. What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students? (RQ4)

Additionally I intended to further explore how teachers could be supported to integrate these topics into their classroom programme; and use teacher feedback to assess the teaching resources provided.

8.2 Method

Information on the resources and training used for this study can be found in Chapter 6, section 6.3.

Data type	Number of respondents
Feedback form	33
Interviews	18
Motivation survey	31
All three	14

Table 8.1: Number of participants that data was collected from.

8.2.1 Data collection

In total 40 people participated in this study at some point through 2015-2016. The number of teachers that each category of data was collected from is shown in Table 8.1.

Of the 40 participants: 10 completed only the motivation survey and then did not continue to participate, 5 completed the feedback form only, 7 completed the survey and feedback form only, 4 were interviewed and completed the feedback form, and 14 were interviewed, completed the survey, and completed the feedback form. The methods of data collection are further described in the following sections.

Motivation survey

From June 2016 onwards, all participants were asked to complete a survey about their motivations for joining the study, teaching this material, and their previous experience with these subjects. They were also asked about their personal feelings about teaching this content, what age group they fell into, and approximately how long they had been teaching. This survey was sent to all teachers who were participating at the time, and to those who were recruited later in the year.

The survey contained the following questions:

1. Why are/were you interested in being part of this programme?
2. Do you have previous experience with teaching programming or Computer Science?
Select one from:
 - Yes —Have taught one or both of these as part of my programme.
 - Some —Have taught one-off lessons with one or both of these subjects.
 - None —These subjects and teaching them are new to me.
3. Please enter some words that describe how you feel about being a part of this programme and teaching Computer Science and programming (e.g. excited, anxious,

motivated)

4. In your own words, what you are hoping to get out of this programme?
5. What kind of ongoing support would you like, to make this programme successful for you?
6. What is your age? Select one from:
 - Under 25
 - 25-34
 - 35-44
 - 45-54
 - 55-64
 - 65+
7. How long have you been teaching? Select one from:
 - Less than 2 years
 - 2-4 years
 - 5-9 years
 - 10-14 years
 - 15-24 years
 - 25+ years

Feedback Form

Participants were given a link to an online feedback form to fill out about each of the lessons they conducted. It was available throughout the two year study. Participants were asked to provide their name, the school they conducted the lessons at, and the year group of the students they taught. This allowed feedback for each separate participant and school to be collated, and the results for different year groups could be compared. Occasionally the feedback entries referred to after-school clubs, extension classes, or classes of specifically selected students, rather than a general classroom. These differences were taken into account during analysis. Completing the form was voluntary and participants were given an incentive to do this; each time they filled out the form 5 times they were sent a \$5 coffee voucher.

Participants were asked to complete the following questions. The question text is written in **bold**, text in *italics* refers to the ‘help’ text provided for some questions, and the unformatted text describes the type of question, e.g. multi-choice, text entry.

1. **What did you do/teach?** Text entry. *1-2 sentences is enough, e.g. “Taught ‘if’ and ‘if, else’ statements in Scratch; students used these to make yes/no questions”, or “Taught binary numbers using cards with 1, 2, 4... dots”*
2. **How confident did you feel while teaching this topic?** Select one from:
 - Very unconfident
 - Moderately unconfident
 - Moderately confident
 - Very confident
3. **How challenging/engaging do you think students found it?** Select one from:
 - Far too easy
 - A little too easy
 - A good challenge level
 - A little difficult/frustrating
 - Far too difficult/frustrating
4. **Which concept(s) was the lesson intending to cover?** Multiple-choice, select all that apply. *The most common “Visual Programming Language” is Scratch*
 - Algorithms: What an algorithm is
 - Algorithms: Following step-by-step instructions (following an algorithm)
 - Algorithms: Creating step-by-step instructions (decomposing problems into steps)
 - Algorithms: Test instructions and correct them (explain and/or correct steps in an algorithm)
 - Algorithms: Describe and/or demonstrate two different algorithms for the same problem
 - Algorithms: Compare two different algorithms for the same problem (which is faster/better for a specific problem)
 - Programming: Understand and give step-by-step sequenced instructions (e.g. Turtle instructions, Beebots)
 - Programming: Using a visual programming language to make a simple sequenced program
 - Programming: Using a visual programming language to make a program that includes simple iteration (repetition)

- Programming: Using a visual programming language to make a program that includes selection and iteration
 - Programming: Using a visual programming language to make a program that includes user input, output, and variables
 - Programming: Using a visual programming language to make a program using different data types
 - Data representation: Using two different symbols (i.e. binary numbers) to represent information, such as integers and letters.
 - Data representation: Binary representations of numeric values
 - Data representation: Binary representations of text
 - Data representation: Binary representations of images
 - Data representation: Binary representations of sound
 - Other: Please describe
5. **Which concepts do you think the students understood well in the lesson?**
Text entry, optional question.
6. **Which concepts do you think the students struggled with in the lesson?**
Text entry, optional question.
7. **Please enter any other comments or observations you have.** Text entry, optional question. *For example: How would you rate the quality of the work students produced during the lesson? Would you do this class again? Did you have any “aha” moments? Do you have any questions about the content the lesson was intended to cover?*

Data collection limitations

I was unable to regularly visit the majority of schools, and this, along with primary school teachers' generally high workload, made maintaining regular contact difficult. I had intended on having students complete short pre and post-tests based on Bebras questions. However, this became logistically challenging, due to the large number of participants, some of whom were teaching multiple groups of students, and by new participants joining at different times. After the majority of participants did not have their students complete the pre-test, or began teaching before conducting it, I choose not to proceed with a post-test.

I had also intended to collect information on students' Reading, Writing, and Maths National Standards results at the beginning and end of the time their teacher spent as part of this study. However, as the majority of teachers confronted administrative challenges in obtaining this information from their schools, and the very low proportion of students who returned permission slips for collecting this information (approximately less than 10%), this effort was abandoned.

8.2.2 Interviews and qualitative analysis

The process of interviewing participants, and the qualitative analysis process used for the interview transcripts and feedback form, are described in Chapter 6, section 6.4.2.

Computational Thinking Coding

I took a deductive approach in my analysis when examining CT skills. The CT skills coded in this analysis were the concepts and approaches from the Barefoot Computing definition of CT. The concepts were *Abstraction*, *Algorithms and Algorithmic Thinking*, *Decomposition*, *Evaluation*, *Logic*, and *Patterns and Generalisation*. The approaches were *Collaborating*, *Creating*, *Debugging*, *Persevering*, and *Tinkering*. Here I will describe each of these concepts and approaches and the criteria used for tagging sections of text with these codes. These are based on the descriptions and examples of each concept in the CAS Computational Thinking Guide for Teachers [49], and the descriptions and examples of each concept and approach on the Barefoot Computing website [11].

Abstraction

“Abstraction makes problems or systems easier to think about. Abstraction is the process of making an artefact more understandable through reducing the unnecessary detail... The skill in abstraction is in choosing the right detail to hide so that the problem becomes easier, without losing anything that is important. A key part of it is in choosing a good representation of a system. Different representations make different things easy to do.” [49]

The criteria used for tagging teachers' comments as showing evidence of Abstraction were:

- Identifying which elements of a problem or situation are important and discounting unnecessary detail.

- Choosing a way to represent something which simplifies it, i.e. create an abstraction of it.
- Programming a simulation of a real-world system.

Algorithms and Algorithmic Thinking

“Algorithmic thinking is a way of getting to a solution through a clear definition of the steps. Some problems are one-off; they are solved, solutions are applied, and the next one is tackled. Algorithmic thinking needs to kick in when similar problems have to be solved over and over again. They do not have to be thought through anew every time. A solution that works every time is needed.

Learning algorithms for doing multiplication or division at school is an example. Algorithmic thinking is the ability to think in terms of sequences and rules as a way of solving problems or understanding situations. It is a core skill that pupils develop when they learn to write their own computer programs.”[49]

The criteria used for tagging teachers’ comments as showing evidence of Algorithms were:

- Creating and writing instructions for turtle-style movement, e.g. KidBots or Beebots.
- Demonstrating and following algorithms, e.g. counting in binary, converting numbers or text into binary, performing the parity trick, and sorting and searching algorithms.
- Planning, creating, and writing programs, e.g. using ScratchJr or Scratch. This required evidence that students were coming up with an algorithm for the program to follow, rather than just evidence of programming by following pre-written instructions or exploring with no particular focus or goal.
- Learning about, and then demonstrating understanding of the general concept of algorithms, what they are, why they are important etc.
- Descriptions of students approaching problems by breaking them down into a set of sequential steps and following these.
- Descriptions of students debugging systematically by checking if they have implemented or carried out an algorithm correctly, or if their algorithm does what they expect it too.

Decomposition

“Decomposition is a way of thinking about artefacts in terms of their component parts. The parts can then be understood, solved, developed and evaluated separately. This makes complex

problems easier to solve, novel situations better understood and large systems easier to design.

For example, making breakfast can be broken down, or decomposed, into separate activities such as make toast; make tea; boil egg; etc. Each of these, in turn, might also be broken down into a set of steps. Through decomposition of the original task each part can be developed and integrated later in the process. Consider developing a game: different people can design and create the different levels independently, provided that key aspects are agreed in advance. A simple arcade level might also be decomposed into several parts, such as the life-like motion of a character, scrolling the background and setting the rules about how characters interact.”[49]

The criteria used for tagging teachers’ comments as showing evidence of Decomposition were:

- Creating and writing instructions for turtle-style movement, e.g. KidBots or Beebots.
- Planning, creating, and writing programs, e.g. using ScratchJr or Scratch.
- Any descriptions of students breaking problems, instructions, or sequences down into smaller steps, or smaller problems.
- Descriptions of students being able to see, and understand, how problems or sequences have been broken down into smaller parts, and why.
- Descriptions of students debugging programs or algorithms when they include students stepping through the algorithm or program and examining each step one-by-one.

Evaluation

“Evaluation is the process of ensuring that a solution, whether an algorithm, system, or process, is a good one: that it is fit for purpose. Various properties of solutions need to be evaluated. Are they correct? Are they fast enough? Do they use resources economically? Are they easy for people to use? Do they promote an appropriate experience? Trade-offs need to be made, as there is rarely a single ideal solution for all situations. There is a specific and often extreme focus on attention to detail in evaluation based on computational thinking... Criteria, heuristics and user needs enable judgements to be made systematically and rigorously.”[49]

The criteria used for tagging teachers’ comments as showing evidence of Evaluation were:

- Students giving, receiving, or acting on feedback when creating digital artefacts, programs, or an algorithm to solve a problem.
- Students testing and improving their work when creating programs, or an algorithm to solve a problem.

- Students identifying that their program had not performed the way they expected it too, or didn't meet the required criteria.
- Students comparing different algorithms or solutions and identifying which was the best for the given situation.
- Learning about, and displaying understanding of, comparing algorithms based on their efficiency and suitability for different situations.
- Evaluating if a solution (program or algorithm) is fit for purpose, e.g. does it meet the users needs? Could it be improved? Does it meet the design criteria?
- Making judgements in an objective and systematic way when creating algorithms and digital artefacts.

Logic

"Logical reasoning enables pupils to make sense of things by analysing and checking facts through thinking clearly and precisely. It allows pupils to draw on their own knowledge and internal models to make and verify predictions and to draw conclusions. It is used extensively by pupils when they test, debug, and correct algorithms. Logical reasoning is the novel application of the other computational thinking concepts to solve problems..."

Logical reasoning is key in allowing pupils to debug their code. They can work with peers to evaluate each other's code, to isolate bugs, and to suggest fixes. During this process, they may have opportunities to employ abstraction, evaluation, and algorithmic design. The novel use in correcting mistakes in code requires logical reasoning."[49]

The criteria used for tagging teachers' comments as showing evidence of Logic were:

- Creating and writing instructions for turtle-style movement, e.g. KidBots and Beebots, as this requires predicting where the 'bot' will move as it carries out the instructions.
- Identifying, locating, and fixing bugs in programs and instructions, unless the description says students debugging method was random and not thought out.
- Descriptions of students developing or displaying logical reasoning, or logical thinking.
- Developing new knowledge from experimenting and testing things.
- Tracing the steps of an algorithm or program and predicting what it will do.
- Students saying phrases such as "If I do this then it will make this happen" and "If this part is right then it means this part must be wrong".
- Identifying patterns and rules in how things work, and using these to make predictions e.g. "They were very quick to discover the patterns, predicting the next card".

Patterns and Generalisation

Generalisation is associated with identifying patterns, similarities and connections, and exploiting those features. It is a way of quickly solving new problems based on previous solutions to problems, and building on prior experience. Asking questions such as ‘Is this similar to a problem I’ve already solved?’ and ‘How is it different?’ are important here, as is the process of recognising patterns both in the data being used and the processes/strategies being used. Algorithms that solve some specific problems can be adapted to solve a whole class of similar problems. Then whenever a problem of that class is encountered, the general solution can be applied.[49]

The criteria used for tagging teachers’ comments as showing evidence of Patterns and Generalisation were:

- Students recognising, understanding, or using patterns.
- Students applying a solution, pattern, or method they had used in one situation to a new situation/context.
- Students recognising when a problem was similar to one they had previously solved, and using strategies from this past problem to approach the new one.

Collaborating

“Collaborating means working with others and it frequently achieves the best results. Teachers plan together and observe one another to develop good practice. Collaboration motivates us to persevere with tasks which might otherwise seem too confusing or difficult. Computer scientists and software engineers often draw on or improve upon others’ work and coding, and this is facilitated greatly in open-source software.”[11]

The criteria used for tagging teachers’ comments as showing evidence of Collaborating were:

- Students working together to create a program, write an algorithm, solve a problem, or achieve a goal.
- Pair programming.
- Students teaching each other or asking their peers for help.
- Students listening to each other, considering others ideas and feelings, and taking turns.
- Descriptions of students social and leadership skills improving.

- Descriptions of teamwork and students valuing teamwork.
- Students recognising each others different strengths and working to support each other.

Creating

“Creating is about planning and making things. Some endeavours involve various media each providing an outlet for creative expression. Software and digital media allow scope for creativity and, by mastering software tools and digital devices, we develop confidence, competence and independence which we can use playfully, experimentally and purposefully in the expression of our ideas and insights.

Programming is itself a creative process. We have ideas about what we’d like to make or solve, analyse the problem, design, write and debug the requisite code and evaluate what we’ve created.”[11]

The criteria used for tagging teachers’ comments as showing evidence of Creating were:

- Creating digital artefacts such as programs, games, artwork, or animations.
- Creating physical artefacts involving computational knowledge or concepts, e.g. binary art.
- Planning, designing, and creating something to meet requirements or solve a specific problem.
- Descriptions of students displaying and developing creative thinking skills.
- Working on creative projects that also involved algorithmic thinking e.g. creating a maze and navigating a Beebot around it, creating a set of instructions for a dance or for telling a story.

Debugging

“Bugs are errors in algorithms and code. Debugging is the process of finding and fixing these and it can often take much longer than writing the code in the first place. There can be errors in logic and syntax. We can think of errors in logic as parts of a story where the plot doesn’t make sense, and errors in coding and syntax as poor spelling, punctuation and grammar. When debugging, it can be useful to display the contents of any variables in our program, to help us understand what’s going on as it runs.”[11]

The criteria used for tagging teacher’s comments as showing evidence of Debugging were:

- Identifying and fixing bugs in programs.

- Identifying when something has gone wrong while carrying out a given algorithm, and determining why this happened, e.g. why the output from using the sorting network was not in sorted order.
- Identifying if a problem that has occurred was due to their thinking and the algorithm used, or if it was due to an implementation error e.g. implementation problem in their code, or a student did not follow the instructions given to them in the KidBots activity correctly.

Persevering

“Persevering is being determined, resilient, tenacious – never giving up... Computer programming is hard. This is part of its appeal – writing elegant and effective code is an intellectual challenge requiring not only an understanding of the ideas of the algorithms being coded and of the programming language you’re working in, but also a willingness to persevere with something that’s often quite difficult and sometimes very frustrating.”[11]

The criteria used for tagging teacher’s comments as showing evidence of Persevering were:

- Teachers stating that their students demonstrated persistence, or their usual persistence levels improved.
- Teachers describing students displaying patience, not giving up when they were working on problems, or working hard even when things took a long time.
- Students displaying motivation and tenacity, taking initiative, and directing their own learning.
- Students adopting a growth mindset. This did not include cases where teachers stated their students already generally displayed growth mindsets before starting this material.

Tinkering

“We often try out something new to discover what it does and how it works: this is tinkering. It’s closely associated with logical reasoning. Pupils build up experiences of cause and effect: ‘If I move this, then this happens.’ It’s a big part of independent learning, without your lead. For young children, it’s the vital play-based experimentation phase, full of questions and surprises. Ideas which seem wrong can be tried, just to see what happens.

For older individuals, tinkering is more-purposeful exploration and making, often through trial and improvement. It helps us to see our use of technology as being about developing

our own understanding, rather than getting a ‘right’ answer; we may be able to do things in many ways.”[11]

The criteria used for tagging teacher’s comments as showing evidence of Tinkering were:

- Playing, e.g. creating ScratchJr programs for fun, or playing with Bee-Bots.
- Not being afraid to try things out and find out what happens.
- Students explicitly using trial and error.
- All cases where teachers described having their students play with something, rather than giving specific tasks.

8.3 Results and Discussion

Motivation survey

In total 31 participants completed the motivation survey. The response from one participant was discarded as they had only answered one question and did not further participate in the study, leaving 30 useable responses. 21 of the participants who completed the survey continued to take part in the study. Of the remaining nine participants, seven informed me soon after completing the survey that, due to external factors, they would no longer be able to participate in the study, or would instead be supporting a fellow teacher who was participating. The other two participants who did not make further submissions did not officially withdraw from the study, but their responses in the survey made it clear they were highly anxious about taking part. Their responses to the motivation survey are further discussed in section 8.3.2.

Feedback form

In total 137 unique feedback entries were submitted, by 33 teachers from 25 different schools. This included one ‘virtual school’, where the class was taught online and was made up of a small group of students from several different schools. A small number of entries did not contain feedback on lessons. Instead, the respondent entered other information they thought would be of interest, such as notes about misconceptions they had when preparing for lessons. These insights were useful, but these entries were not included in the analysis of the feedback. Some responses covered several lessons that took place over multiple days. In these cases, they were divided into separate entries. Responses that focused on topics that

were not relevant to CS, programming, or CT, such as lessons about using computers, were discarded. The useable feedback data came from 32 teachers, and amassed to 112 entries for general classroom environments, 18 for non-typical environments or target groups of students, and one for a workshop which was conducted with students from multiple schools.

Interviews

18 teachers were interviewed at the end of 2016. The core themes from the thematic analysis of the interviews and feedback entries were:

- Computational Thinking skills
- problem solving
- Key Competencies
- Cross-curricular learning and learning transfer
- Individual student impacts

These themes, the lessons covered, and information on the participants and their schools, are discussed in this section.

8.3.1 School information

This study included Public (State), State-integrated, and Private schools. The decile ratings of the schools involved were obtained from the Ministry of Education website, where the deciles of all NZ schools are publicly available. The distribution of school deciles, and the categories of schools are shown in Tables 8.2 and 8.3. This study covered a range of deciles; however, as by definition each decile represents 10% of NZ schools, decile 10 schools were particularly over-represented in this study.

In both 2015 and 2016 approximately 83.1% of NZ schools were State schools, 13.1% were State-integrated, and 3.4% were Private.² With 9.5% of schools in this study being Private, and 9.5% being State-integrated, these two types of schools were over and under-represented, respectively.

² <https://www.educationcounts.govt.nz/statistics/schooling/number-of-schools>

Decile	Number of Schools	Percentage
1	0	0%
2	0	0%
3	2	9.5%
4	1	4.8%
5	2	9.5%
6	2	9.5%
7	0	0%
8	3	14.3%
9	2	9.5%
10	9	42.9%

Table 8.2: Deciles of schools reported on in 2015 and 2016.

School Type	Number of Schools	Percentage	Decile range
Co-Educational, State	17	81.0%	3 to 10
Co-Educational, State-integrated	2	9.5%	8 to 10
Single-sex (Female), Private	1	4.8%	10
Single-sex (Male), Private	1	4.8%	10

Table 8.3: Types of schools reported on in 2015 and 2016

8.3.2 *Participants*

The vast majority of participants were employed as full or part-time primary school teachers. Those who were not teachers were volunteers who provided supplementary lessons and teacher support at a primary school they were not usually employed at. As participants were self-selected, the results of this study are potentially biased towards teachers who are more confident with, or passionate about these subjects than most NZ primary school teachers.

There were significantly more teachers in the 45+ age groups than younger groups, and with 15+ years of teaching experience. There were no teachers in the 65+ age group. The distribution of ages and years of teaching experience are shown in Tables 8.4 and 8.5 respectively. The 45-54 year old age group was overrepresented in this study when compared

Age group	Under 25	25-34	35-44	45-54	55-64
Number of teachers	1	0	6	11	3
Percentage	4.8%	0%	28.6%	52.4%	14.3%

Table 8.4: Participant age groups.

Years spent teaching	2-4	5-9	10-14	15-24	25+
Number of teachers	3	1	2	10	7

Table 8.5: Number of years each participant had been working as a school teacher

to the age distribution of all NZ teachers, 24.9% of whom are in this age group. 18.6% of teachers in NZ are in the 25-34 age group and, as there were no teachers in this age group in my study, teachers under the age of 34 were highly under-represented. This is interesting considering the common stereotype that computing related topics and computer usage are easier for, and more appealing to, the younger generation of teachers. Twenty-six of the participants were women, and only four were men. This is relatively unsurprising considering the large gender bias towards women in primary school education in NZ, where over three-quarters of teachers identify as women.

While the support and training available were the same for all participants, there was a huge variation in how much each used these, and which parts they used. Some teachers had prior experience with these topics and so did not attend as many workshops. There were teachers completely new to these topics and very nervous about trying them, who made use of as much support as they could. Other teachers new to this topic did the opposite, and after just one session of training continued with no support. Some did this as they quickly felt comfortable and confident with the material, but some because they taught a small number of lessons and did not continue.

Participants were asked about their confidence levels in the motivation survey, the feedback form they completed during the programme, and in the interviews. They were also asked about their motivations for joining this programme in the survey and interviews. This allowed for a comparison between teachers' feelings before, during, and after the programme.

The most common themes in the motivation survey responses were feelings of determination and excitement, frequently alongside general anxiety and apprehension. As would be

Amount of experience	Number of teachers	Percentage
Yes —Have taught one or both of these as part of my program	5	23.8%
Some —Have taught one-off lessons with one or both of these subjects	10	47.6%
None —These subjects and teaching them are new to me	6	28.6%

Table 8.6: Number of years each participant had been working as a school teacher

expected, there was a clear difference in confidence levels between teachers with different amounts of experience with these subjects. Teachers with no prior experience overwhelmingly reported feelings of trepidation, while those with experience did not. Two of the teachers who reported high levels of anxiety and self doubt did not go on to provide feedback or be interviewed. One described themselves as feeling “overwhelmed”, and indicated the ideal type of support for them would be to have someone more experienced in these topics help them teach the first few lessons to get them started. The other said they felt “confused” about where to start. They wrote that model lesson plans would be the most useful type of support. Instructions were provided for all CS Unplugged activities, however, at the time these resources were not developed into full lesson plans. Full lesson plans were available the following year and may have assisted this teacher to continue teaching these subjects.

All of the 21 teachers who continued with the study reported some type of positive feeling about being a part of the programme. 19 described themselves as ‘excited’ or ‘motivated’, while the others were ‘curious and interested’, and glad that they were going to be a part of the study. The majority of these teachers were relatively confident, and had some degree of prior experience with these topics, as shown in table 8.6. three of the five teachers who answered “yes” to prior experience had been involved in the study for at least a year before completing the survey. The large number of teachers with some degree of prior experience with these subjects was to be expected, as the majority of participants found out about the study through the CS4PS workshops (described in Chapter 6, section 6.3.4).

Participants frequently cited their students as their motivation for taking part in this programme. They saw the importance of this subject for their students’ futures and wanted

to do the best they could for them. They wanted their students to be engaged with the topic, excited about it, and excited about their learning in general.

- “I want to be able to ensure I am doing the best for my students and give them skills that are relevant to their digital world.”
- “I believe that I can help prepare our students for a world where the internet of things is a reality. By exposing them to computational thinking at a level appropriate for their age they will gain a better understanding of what makes their world work. They will be better able to make an informed decision about whether this is an area they would like to pursue in later studies.”
- “I believe that it is essential for young people to learn how to control and create technology. Understanding how computers work and being able to program them is the most powerful enabler I know in the 21st Century.”
- “I want to have some fun, engaging classroom activities that make my maths programme more hands on and interactive and broaden their learning focus”
- “Student engagement and enjoyment in something new.”

Many also wished to up-skill themselves in this area, and become more comfortable with the topic of Digital Technologies.

- “I know this is a skill I need to learn and this opportunity will give me lots of support and motivation”
- “I am interested because, as part of my professional development I would like to be less fearful of these technologies.”
- “To upskill on my teaching practice.”

Supporting other staff within their school was a motivating factor for several teachers. These teachers were generally the lead teacher (officially, or unofficially in some cases) for ICT or DT in their school, and were responsible for getting other teachers in their school on board. Conversely, other teachers were hoping that over the course of the programme they

would gain the support of more staff at their school, their students' parents and caregivers, and teachers at other schools.

Despite these many motivating factors and general excitement teachers felt, there were many who were 'stressed' about the amount of work they felt this would be for them. Over half of the participants described these types of feelings. Two teachers highlighted the difficulty of finding the time to fit this subject in:

- "I thought it would be great to get some PD, and hopefully a motivator to do something in class about it...as it is not really on the priority list with the million other things we have got going on."
- They were hoping this program would help them "to include aspects of the programme into existing frameworks - not an add on. I won't fit it in otherwise."

Time, workload, and the already very full primary school curriculum were the most common struggles teachers described in the interviews. 16 of the 18 interviewees identified this as one of their main challenges. However, most of these teachers said that this was a problem for primary school teachers in general, and not unique to this subject. The high workload for NZ primary teachers has been identified as a national concern, and a large contributor to recent teacher strikes and teachers leaving the profession [70, 71].

Teachers' levels of confidence reported in the feedback form were surprisingly high, considering that teaching these topics was still relatively new for the vast majority of them. The number of times each level of confidence was reported is shown in Table 8.7. This was potentially biased as teachers with higher confidence were, presumably, likely to teach more classes than those with low confidence. They would, therefore, have been more likely to fill out the feedback form multiple times. To show a less biased example the mode response for each teacher was taken. In cases where a mode could not be taken the confidence level of the participant's most recent entry was used. This is shown in Table 8.8. The distribution of this data was however similar to that shown in Table 8.7.

The interview transcripts showed that teachers tended to feel less confident and competent when it came to programming, and several teachers said it was something they were afraid of. They were much more likely to feel comfortable with the CS Unplugged lessons. The vast majority felt that they, at least partially, understood the concepts behind the CS Unplugged lessons after teaching them several times. This progression of quotes, all from

Response	Number of responses
Very unconfident	9
Moderately unconfident	15
Moderately confident	66
Very confident	45

Table 8.7: Teacher responses to the question “How confident did you feel while teaching this topic?”

Response	Number of teachers
Very unconfident	2
Moderately unconfident	3
Moderately confident	23
Very confident	5

Table 8.8: Most common response of each participant

the same interview, illustrate this teacher’s feelings, and how their confidence changed over time:

- “I feel like I am quite computer literate, although I certainly can’t do any computer programming whatsoever, it actually terrifies me.”
- “When I went to the very first session [on CS Unplugged] I actually thought oh my goodness I know nothing! I thought I can’t do any of that, I don’t even know what those things mean! It was quite terrifying. But then when I downloaded the book and focused less on the bigger picture and more on the activities, then I felt like I was getting more headway in my own mind, and then that led to deeper understanding of the bigger picture.”
- “I personally don’t feel very confident on Scratch and I have older kids next year, so I will learn how to use Scratch, but that’s probably the area I feel weakest in. The other activities I feel very confident with those.”
- “I think my knowledge has increased.”

Some teachers became much more confident in their programming ability, though none felt as though they were ahead of all of their students. This however did not concern the teachers who commented on their students' more advanced abilities because, as one stated, "that's great, because I'm only here to get them started and interested". Others reported that they gained no knowledge of programming during the time they taught it, or did not engage with it at all. In these situations the teachers either did not teach any programming, or instead gave students resources they could use independently, such as Code Club projects, and had their students work through these by themselves and assist each other. Being completely new to these subjects was not seen as a negative by all teachers, and some found it to be a positive for both them and their students. One participant commented happily:

- "Honestly, I had no idea what I was doing when I first started doing it with the kids. I just said 'let's play with this!', and they'd ask 'how do you do that?', 'I don't know! Remember I'm learning too!' So, I would have it up on the board and I'd be trying it at the same time as them".

There was no stress in their voice when they said this. Instead, there was a lot of excitement. This encouraged students to be more persistent and self-reliant in several classes. Many teachers felt that learning alongside their students, and their students being aware of this, was beneficial. In contrast, other teachers felt that their lack of knowledge and confidence with the subject material meant they were doing a disservice to their students, especially when they found they could not answer students' questions. Despite this, even this group of very unconfident teachers were glad they had pushed themselves to try teaching these subjects. During the interviews, I asked each teacher if they were glad they had taken part in this, or if they would be teaching it again in the future. The answer to one or both of these questions was a firm yes from every interviewee.

- "Yes, absolutely. It is part of the future, well actually it's not the future, it's part of what we should be doing now'."
- "Absolutely delighted, really loved it, been really passionate about it. Its been brilliant."
- "Definitely. I was definitely nervous before I started and was thinking I don't know if this is for me. But I'm really glad that I started and now I feel like I am able to help other teachers get into it."

All stated that they would be teaching this topic in the future, in some form. The vast majority were planning on using the lessons again, wanted more resources to extend their teaching in the future, and planned on attending future CS4PS workshops. One teacher did not say definitively whether or not they were pleased they had taken part. They had taught the material as part of an after-school club, rather than during class time. They felt that this had been a mistake and had made the classes less successful. In every interview, teachers expressed overwhelmingly positive feelings towards teaching these topics, despite the challenges of time, workload, stress, and how much prior experience they had with these subjects. This complete consensus was very surprising, and also encouraging for the future of CS and CT curriculum.

8.3.3 *Lessons taught*

The feedback gathered covered many different lessons, conducted with a wide range of age groups. The majority were about lessons that took place in usual class time with a typical class of students. The remaining entries covered after-school clubs, outreach workshops with students from multiple schools, a virtual classroom, and classes with specifically selected groups of students. These groups with selected students included extension classes, lessons with a group of ‘gifted and talented’ students, and students who were disengaged and struggling with their learning. This also included one group of students who were chosen by their usual classroom teacher to take part in a programme of lessons because they thought the students would benefit from extension in this area but were not a usual extension class. This programme of lessons was taught by a volunteer who worked in the field of e-learning, rather than the students’ usual teacher. They were chosen at their teacher’s discretion, and not based on their abilities in other subjects.

Table 8.9 shows which lessons were taught according to the feedback form, and the range of age groups they were done with. The number of times each concept was taught in a lesson, as reported in question four of the feedback form (options listed in section 8.2.1), is shown in Table 8.10. As would be expected, many responses reported multiple concepts in each lesson. Therefore, the total number of concepts is much higher than the number of feedback responses.

The general consensus among teachers was that each of the topics covered could be taught in a suitable way for each age group they taught, and their students’ responses to the lessons were overwhelmingly positive. The fact that these lessons worked for different age groups

Activity	Year levels	Computational Thinking links
Programming:		
Bee-bots and similar	0/1 - 8	Algorithms, Decomposition, Evaluation, Logical thinking
ScratchJr and similar level programming environments	3 - 6	Algorithms, Decomposition, Evaluation, Logical thinking
Scratch and similar level programming environments	3 - 8	Algorithms, Decomposition, Evaluation, Logical thinking
CS Unplugged:		
Unplugged Programming: Kidbots, mazes etc.	1 - 8	Algorithms, Decomposition, Evaluation, Logical thinking
Binary numbers: number representation	1 - 8	Algorithms, Decomposition, Patterns
Binary numbers: representation of text, images, or sound etc.	2 - 8	Abstraction, Algorithms, Decomposition, Patterns
Error detection and correction: parity and barcode magic	4 - 8	Abstraction, Algorithms, Patterns
Searching Algorithms	3 - 8	Algorithms, Decomposition, Evaluation
Networks: The muddy city	3 - 6	Abstraction, Algorithms, Evaluation, Logical thinking
Routing and deadlock: The orange game	3 - 8	Algorithms, Decomposition, Evaluation, Logical thinking
Sorting networks	3 - 6	Algorithms, Evaluation
Information theory and decision trees	5 - 6	Abstraction, Evaluation, Logical thinking, Patterns
Sorting Algorithms	5 - 6	Algorithms, Evaluation

Table 8.9: Activities that were reported on in the feedback form responses

Concept taught, as reported by participants	Times reported
Algorithms	
What an algorithm is	23
Following step-by-step instructions	31
Creating step-by-step instructions	36
Test instructions and correct them	31
Describe and/or demonstrate two different algorithms for the same problem	14
Compare two different algorithms for the same problem	13
Data representation	
Binary representations of numeric values	26
Binary representations of text	14
Using two different symbols to represent information	10
Binary representations of images	8
Binary representations of sound	4
Programming	
Understand and give step-by-step sequenced instructions	33
Using a visual programming language to make a simple sequenced program	30
Using a visual programming language to make a program that includes input output and variables	15
Using a visual programming language to make a program that includes simple iteration	20
Using a visual programming language to make a program that includes selection and iteration	15

Table 8.10: Total number of times each concept was reported in the feedback form.

Response	Number of responses
Far too easy	0
A little too easy	2
A good challenge level	119
A little difficult/frustrating	10
Far too difficult/frustrating	0

Table 8.11: Teacher responses to the question “How challenging/engaging do you think students found it?”

was likely influenced by teachers’ expertise. Their pedagogical skills would have helped them to modify the content themselves, to suit the age level of their students. For example, when interviewed one of the teachers commented that after teaching the lessons several times, they “saw ways to take some of the lessons that were in Unplugged, and develop them again specifically around this age, expanding them, or making them more hands-on for the kids”.

The form asked teachers to rate how challenging lessons were for their students. The number of times each challenge level was reported in the feedback form is shown in Table 8.11. 10 lessons were identified as “A little difficult/frustrating”. However, four of the pieces of feedback stated the difficult part of the lesson was unrelated to the CS, CT, or programming concepts being taught, but to issues such as lack of access to resources, or children not knowing which directions were left and right. The six remaining feedback entries, identified as “A little difficult/frustrating” came from three separate teachers. Four of these entries were from the same teacher, with one each coming from the other two teachers. These six lessons covered year levels from 2-8, and so the difficulty does not appear to be related to the age group being taught. In fact, two of these entries reported on the same lesson being taught to different age groups (but using different, age-appropriate, programming languages). These entries stated that both groups struggled with exactly the same CT concepts. Seven of the total eight lessons were on programming concepts and algorithmic thinking for programming, and one was on the ‘Image representation’ CS Unplugged activity.

The main challenges in these lessons were all very similar. Students struggled with working methodically, and in following step by step instructions sequentially. For example, teachers commented that students struggled with the following tasks :

- “The process of debugging - most wanted to randomly remove blocks in the hope it

would work.”

- “Planning, thinking through the steps in a logical and systematic approach to then be able to then write their program.”
- “The level of accuracy was surprisingly low. They tended to rush. Those who systematically crossed out each number and row made fantastic progress.”

However, in each of these lessons, the students were able to eventually overcome these difficulties. For example, in a lesson where students were having difficulty constructing their programs the teacher found that “The struggle was short-lived. As soon as one student had a breakthrough, then they were encouraged to show others and success spread like wildfire”. The difficulty level of these lessons did not appear to be related to the teachers’ level of confidence.

Each of the 18 teachers interviewed was asked if they thought these lessons had a positive impact on their students. 17 of the teachers reported that yes, the lessons had been beneficial. The teacher who did not report the lessons were beneficial stated that they could not make a judgement on this, because they had taught after school, outside of the students’ normal classes, and did not work with this group of students during the day. This meant they were not able to observe their behaviour and work in their usual classes and classroom environments, so were unable to observe any potential development the students may have displayed.

They were also asked if there had been any negative impacts on students. 15 of the teachers stated they had not observed any negative impacts. This may appear to indicate that three of the teachers *did* observe negative impacts, but their answers were not as simple as that. Interestingly, they all described the same observations, despite being in different schools and not working with each other during this programme. They said the only negative impact they could think of was some of their students became frustrated or upset when they first began learning the material, as they found it challenging — when asked what negative impacts there were their answers were:

- “Only for my really really high functioning kids,... They were just frustrated really easily because it didn’t work the first time. It was the high functioning kids that were like that.”

- “Maybe a little frustration. Like, when the binary numbers or the network wouldn’t work out at the end they would go *‘oh this is stupid’*. But to get them to go back and say *‘let’s find out what went wrong and let’s start again and see where we went wrong’*, and solve it that way and have success at the end. So just that being frustrated[sic]. It wasn’t a negative effect, but that reaction from the kids who think *‘I’m clever and I can do this’* and then it didn’t work and they would say *‘oh this is stupid’* or something. So, getting them to try and do it again... It was that persistence again and not giving up.”
- “Other than just the short-term frustration for those kids, but that’s nothing that we wouldn’t see in any lessons with some kids, there’s always going to be some kids with that struggle, and basically once they can persevere and show that growth mindset they can actually push through and once they collect they get a lot out of it.”

While a range of students in each case displayed these feelings, they were particularly common among the higher achieving students in the classes. The teachers suspected this was because the material put these students on a more even footing with their classmates, as it was a new topic for all of them. They seemed frustrated that they were not able to get the correct answers as quickly as they may have been used to doing in other classes. One teacher said they did not think this behaviour was in any way unique to this particular subject, and they observed it often when these students encountered something new. All three of these teachers stated this frustration was temporary. Two remarked that this could be interpreted as a positive impact, as it gave students an opportunity to practice diligence and persistence.

There were no references to any negative impacts on students in the feedback form responses, and almost all entries described one or more positive impacts.

8.3.4 Computational Thinking

All 32 teachers observed, or described, students carrying out or exercising at least one CT skill. Unsurprisingly, there was a large difference in the number of concepts and approaches teachers observed based on whether or not they had been interviewed, and the number of times they submitted feedback. Two teachers submitted significantly more feedback than others, one of whom was Wendy. Between the feedback and the interviews, all concepts and approaches were reported 40 or more times, other than Abstraction which was reported 6

times, far fewer than the others. For descriptions of these concepts and approaches, and the criteria for tagging these, please refer to section 8.2.2. The total number of times these were each identified are shown in Table 8.12, and in Figure 8.1. The number of teachers who observed each of the concepts and approaches is shown in Table 8.13.

CT Skill	Number of observations:		
	Feedback	Interviews	Total
Concepts:			
Abstraction	3	3	6
Algorithms	123	91	214
Decomposition	47	85	132
Evaluation	22	43	65
Logic	33	62	95
Patterns and Generalisation	19	28	47
Approaches:			
Collaborating	27	55	82
Creating	27	65	92
Debugging	22	61	83
Persevering	2	38	40
Tinkering	11	38	49

Table 8.12: The Computational Thinking Concepts and Approaches, which were identified in the feedback, and interview transcripts.

Each CT skill, other than Algorithms, was much less likely to be observed by teachers who only provided feedback and were not interviewed. This is unsurprising as the interviews provided much more detailed observations. One of the quotes that stood out the most, on the topic of CT, was from Angela, as she reflected on the three years she had been teaching CS and programming:

- “When you can actually see how differently students think, you know that they’re really learning to program. They’re thinking of the algorithm... That’s Computational Thinking, to be approaching the problem the way they do.”

In her interview, she talked about how both her students’, and her own thinking had changed, and how their problem solving skills had evolved. As she had been teaching this material for the longest and had observed groups of students over consecutive years, she

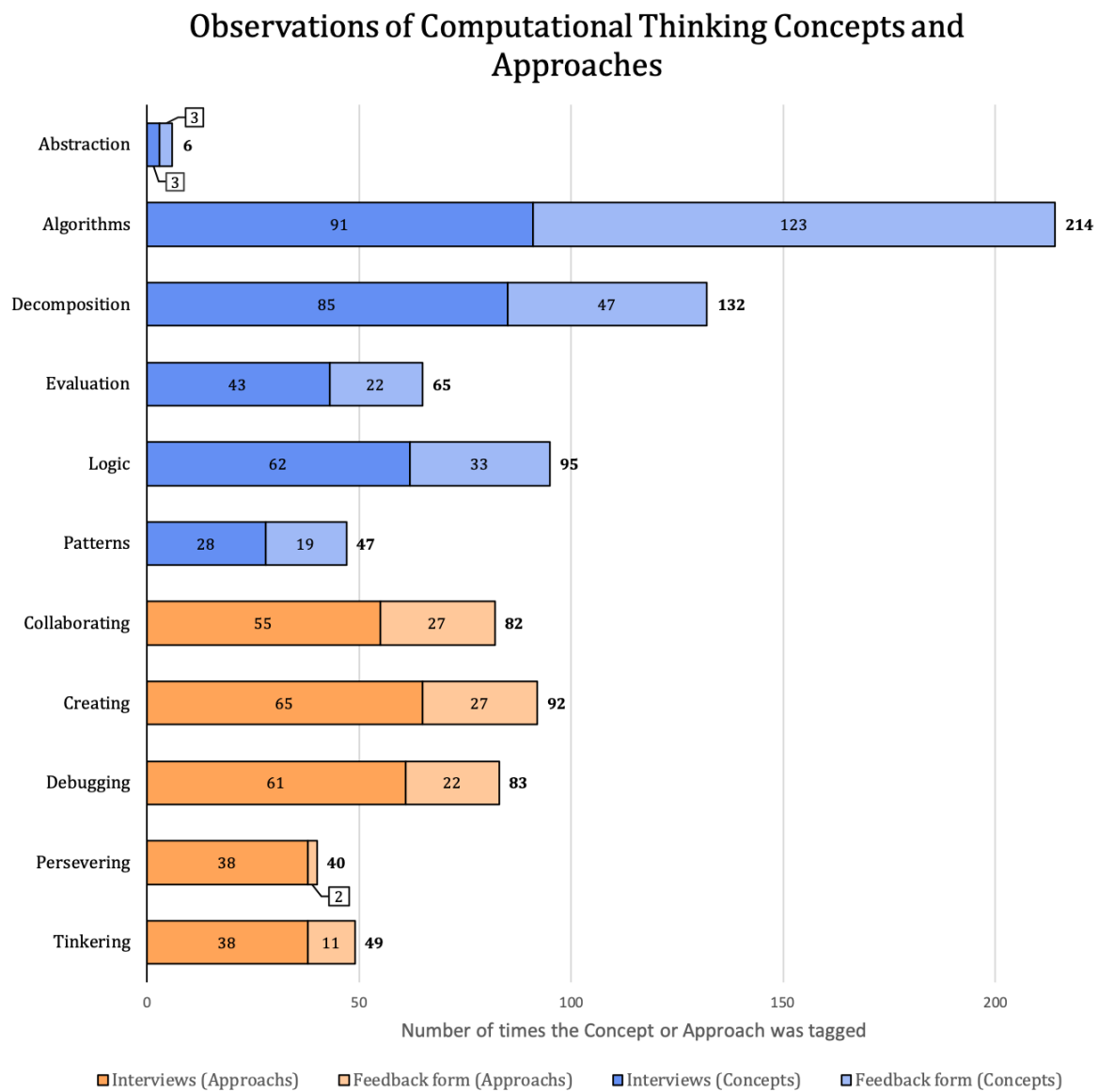


Figure 8.1: The Computational Thinking Concepts and Approaches, which were identified in the feedback, and interview transcripts.

CT skill	Gave feedback only, out of 14		Gave feedback and interviewed, out of 18		All teachers, out of 32	
Concepts						
Abstraction	0	0%	3	17%	3	9%
Algorithms	14	100%	18	100%	32	100%
Decomposition	6	43%	17	94%	23	72%
Evaluation	1	7%	16	89%	17	53%
Logic	4	29%	17	94%	21	66%
Patterns	2	14%	14	78%	16	50%
Approaches						
Collaboration	3	21%	14	78%	17	53%
Creating	4	29%	16	89%	20	63%
Debugging	3	21%	18	100%	21	66%
Persevering	0	0%	15	83%	15	47%
Tinkering	1	7%	16	89%	17	53%

Table 8.13: The number of teachers that each Computational Thinking Concept and Approach was observed by.

was better placed than the majority of other participants to comment on how this impacted students over the longer term.

Through the interviews and feedback, there were examples of CT skills being used individually, but the majority of observations involved combinations of these. With the many interconnections between the concepts and approaches, this was not surprising.

Abstraction

Abstraction was observed significantly less often than other CT skills. This is interesting, but not particularly surprising considering the young age group of students being observed. Despite the low number of Abstraction observations, there were three particular cases identified which gave very distinct examples of students performing Abstraction. The first two of these both came from the same teacher.

- *Maths word problems:* One of the examples of Abstraction given in the CAS teachers guide is students using Abstraction when tackling word problems in maths to identify which information is, and is not, needed to solve the problem. While this is an example

of this happening in maths, rather than CS, one teacher discussed how this tied in well with the CS and programming work their students did, and how it enforced this problem solving method in maths. “Because they’re not so great with the word problems ... initially, we were cutting things out so they could put them in the line, after we had done Scratch so we could then put them in a program. So like ‘10 strawberries’, ok cut that out, ‘add’ ok cut that out, put that one over there... so it has helped them move things across with their word problems.”

- *Logic problems:* In their interview, a teacher described how at the beginning of the year they had given their students logic puzzles to work on and then gave them puzzles of the same type at the end of the year. These puzzles required students to use information and rules to match pairs, for example matching different characters to the activities they did based on a set of rules and information about their likes and dislikes. At the beginning of the year students “just couldn’t do it and they were saying ‘but it doesn’t say that character’s name?’, [I said] ‘No but it says that [if] that character is doing that and that has that, then that character must do that’... I had kids in tears”. But at the end of the year, the students not only completed these, but enjoyed them, and the teacher believed this was largely due to the CS and programming lessons. She described how the majority of students now understood that they didn’t need to know all the details of the situation to find the answers, they could use the abstract information they had to do this. This was also a clear example of students developing logical thinking skills.
- *Muddy City extension:* One of the classes who did the Muddy City Unplugged activity extended this by creating new maps and adding new requirements to the challenge. These included deciding the length of each tile and how much each tile would cost. Students had to figure out how to find the shortest routes between different houses, and the cheapest way to pave the whole city while minimising the distance between destinations. This was added as a way of including maths in the activity. Students however soon realised they could simplify this by ignoring the length and cost of each paver, only focus on the number of pavers used, and then calculate the distance and cost at the end of the activity. They identified and discounted the unnecessary details, thus performing Abstraction. This was also a strong example of students performing evaluation (along with other CT skills).

The remaining observations of Abstraction were of students coming up with different

representations they could use to record a set of directional (e.g. Beebot or KidBot) instructions. There were a large number of cases that could be considered to involve Abstraction but were not covered by this criteria. These were lessons and activities related to binary data representation, and cases of students creating programs that could be considered to be simple simulations.

While learning and using different representations of data is an important type of abstraction frequently used in computing, the descriptions of data representation lessons were not considered examples of Abstraction, as students were not choosing these representations themselves. They did not have to consider or choose what representation was most appropriate for the data they were using, or why representing data in these different ways was related to the concept of abstraction or complexity. Simply learning about and using data representations was therefore not considered enough evidence to tag an observation with Abstraction.

Multiple students created programs that could be argued to be simple simulations. For example, one teacher asked their students to do Scratch projects related to their Science Fair project and described one student “who actually designed a lab and you press different buttons and the nails go in the coke, and sometimes one will go rusty and the other will go... whatever [speaker trails off]”. From this information alone it was difficult to determine whether this student created a program in which pressing one button triggered one animation (i.e. the nail going rusty) and one button triggered another, or if they created a program that defined some type of rule about when a nail will go rusty, which could qualify as a simulation. However, this was not considered enough evidence of a simulation of an abstract concept being created, partially as the other student’s projects described in this same section were simple animations.

Algorithms and Algorithmic Thinking

Algorithms and Algorithmic Thinking (referred to from here on as just Algorithms) was the most frequently identified CT concept within the feedback and interviews. This was unsurprising as algorithms was one of the three core topics these lessons were intended to teach, is central to CS and programming, and the data representation activities all involve the use of algorithms as well.

Students were observed learning about or demonstrating this concept through programming and Unplugged programming activities; learning and then carrying out different al-

gorithms for sorting, searching, manipulating binary numbers, error detection, and error correction; articulating an algorithm for someone else to follow; and by walking through algorithms and programs step-by-step when debugging and evaluating these.

Some examples of the tagged sections are:

- When describing students helping their peer with doing the parity trick — “The others could give him a clear strategy on how to do it, it was sequential and how to follow a process rather than just being random.”
- From feedback — “They all understood the binary search algorithm”
- From feedback, concepts students understood well — “Algorithms: how to test then correct them”
- From feedback, concepts students understood well — “That an algorithm is a sequence of instructions that when followed exactly always give the same result. Relating this to programming - if, then, else, repeat”
- When describing benefits to students — “The deeper understanding of sorting, categorising, putting things like the weights into order, and the sorting network, and even the orange game, of trying to follow those in your mind. Following a set of possible solutions, it is so good for them.”
- When describing writing algorithms for Bee-Bots — “That was another thing that we did last year that the kids loved. We were lent the Beebots and made big maps with big pieces of cardboard with squares and they coded the Beebots. That was really interesting, they had these big boards and all these different games that they had made.”

Decomposition

Decomposition was the second most common CT concept and skill identified. It almost always occurred in conjunction with algorithms, which makes sense as creating an algorithm (whether or not it is then implemented in a program) requires students to break a problem or process down into individual steps (must be more than one step). It also frequently occurred with Debugging, as by its nature debugging will generally involve having to examine and walk through individual steps in a program or algorithm.

Students were observed learning about, or demonstrating this concept through programming and Unplugged programming activities; by teaching, or assisting others with algorithms and programming, by breaking these down into steps and explaining these; by articulating

an algorithm for someone else to follow; by walking through algorithms and programs step-by-step when creating, debugging, and evaluating these; by testing and examining individual parts of algorithms and programs, and then building these up. Some examples of the sections tagged as decomposition are:

- Describing students writing Beebot (i.e. turtle-style) instructions for drawing numbers on the ground, or moving a counter from a start to an end point —“it was really good to see the kids that could actually break that number down into steps, and those kids that couldn’t see those smaller steps. So we then had to get them breaking it down and simplifying it. Once they actually got a bit more confident in that then those kids that were getting a wee bit frustrated, they then could see that it was quite an easy task when they were breaking it down into just those smaller chunks.”
- From feedback, concepts students understood well —“Debugging - decomposing the problem to find the problem and then develop a solution.”
- Describing two different students —“they were the one that was really decomposing beautifully, and doing the troubleshooting”, “we’ve got this little year 4 and their decomposition skills were remarkable.”
- When describing the positive impacts of programming —“It’s then bringing out those sorts of concepts, like the computational thinking side of things and being able to break those tasks down.”

There were also a significant number of cases where teachers described students struggling to break problems down when they first encountered them, and in most (but not all) of these situations, they described how students became better at this. For example, one teacher described how students were asked to come up with a project to do or problem they wanted to solve (which they struggled to do independently), but when they tried to start this “the problem would be so big, they wouldn’t know how to even get started”. In their interview, they discussed that with support and time the students became much better at approaching a problem and breaking it down into smaller steps, and to think about it algorithmically.

Evaluation

Evaluation was frequently observed alongside Debugging, Creating, Algorithms, Collaboration, Logic, and Decomposition. Students were observed learning about, or demonstrating this concept through creating programs with a specific goal, such as telling a story, teaching someone a concept, or making a game, individually or as a team; when working as a team

to solve a problem (such as the Muddy City) and testing each others ideas; when learning about searching and sorting algorithms and displaying an understanding of how different algorithms were more efficient than others; and identifying when something had gone wrong whilst carrying out an algorithm or running a program, and working to fix or improve it. Some examples of the sections tagged as evaluation are:

- Discussing ideas with students and observing them understanding this — “We did a lot of talking about how it is actually about solving problems. It is about being accurate and precise and having the right sequence. It’s about seeing where the problem is and going back and fixing it.”
- Discussing benefits of CS Unplugged activities — “it encourages them to think in different ways, they definitely had to use some trial and error for things like the binary activity. Same with that muddy city, that trial and error, give it a go, work out what else you could try.”
- Describing students programming — “the children, they were able to work through and say ‘that’s not working OK how do we make things go the way we want them to?’ ”
- Describing things students learnt — “We taught them how to give critical feedback. They showed us, they verbalized, they worked on it and came back again. If we follow the way in which technology works, which is the planning, the design and getting your stakeholder feedback. They are learning that and getting exposed to it at an early age. The planning part and the testing, and the re-planning and testing again.”
- Describing students projects — “We brought a design thinking element into the program as well, so we use the design thinking framework ... for game creation. In Hopscotch we didn’t want them just playing, creating games without a purpose. We actually wanted them to really start looking at doing a needs analysis, and creating prototypes, and going through that testing stage. They really loved that process... they got that. We had an arcade for a whole session, and they were writing down the positives and the negatives of the games. That was really cool, really enjoyed that.”
- From feedback, concepts students understood well — “We compared a linear search and a binary search and took our time to explore these search techniques with the same sample of data. When we got to finding a number within 1,000,000 and worked out it could be done in 20 guesses, there was much excitement.”
- From feedback on the Muddy City activity — “Some children demonstrated how to swap routes over so the route with the least pavers could be created. They used expressions like minimising each route in order to save pavers and finding the most

efficient route.”

Logic

Students were observed learning about, or demonstrating this concept through writing and debugging algorithms and programs; carrying out algorithms from CS Unplugged activities, identifying when something had gone wrong and why this had happened; playing with, and exploring block-based programming environments; and discussing, creating, and testing algorithms and strategies for solving CS Unplugged problems.

Logic was often identified by teachers as a skill they wanted the CS and programming lessons to develop in their students, and these teachers explicitly stated that their students logical thinking skills had indeed improved. An example of this was given previously when Abstraction was discussed, of students improved performance solving logic puzzles. Some other examples of sections tagged as Logic are:

- Describing what a student said while debugging their program —“That’s the wrong way around, I’ve done and negative when I should have done a positive, I flipped it the wrong way.”
- Comments from multiple teachers on the benefits for students —“I think it’s the problem solving skills, when you look at the logic in it, and the ability to actually see a pattern, or a path, an outcome and those steps to get to it”, “coding has the logic and reasoning the mathematical concepts, the science concepts, it’s all there”, “This is an excellent activity to challenge logical thinking”, “It teaches logic and reasoning, which is really low, unreal low. I was surprised by some of those results with some of those kids.”
- Feedback on students predicting results of a program —“They are developing great spatial awareness and many can programme [sic] the beebot without having to move it from the spot so they can visualise which way it needs to turn”

In the feedback form there were four entries from one teacher describing what students came up with when asked what skills they had to use during their lesson, and what they thought they did well. Students came up with multiple skills and in all cases one of these skills was logical thinking. The same teacher who submitted this feedback also described another lesson which was a good example of students exercising logical thinking. This lesson was not an official CS Unplugged activity, but was an ‘unplugged’ activity on ‘inputs and

outputs' that the teacher created.

They described the lesson in the feedback form as students “acting out inputs and outputs, so that there was a ‘machine’ in the middle of the group (made with children moving) which applied a rule to the input then put it out to the ‘output’ person. Those watching heard the input for each group then their eventual output and had to guess the rule which was applied in the machine”. In their interview they gave the example “this person might pop up and say 5, and then they [the other group] get together... and this person would pop up and say 15, and the other group had to guess what happened inside the machine. So someone would say it was multiplied by 3, and the team will say that’s not what we did, the others would say OK you’ll have to do something else to show what you did. So then they’ll say 6, and the machine pops out 16, and they’ll say oh it must be add 10.”

Patterns and Generalisation

Pattern recognition and generalisation skills were most commonly observed in binary activities, and cases of students applying one problem solving or thinking method to multiple different problems in both programming, and other subjects. The other subject that students and teachers most commonly connected with patterns was maths, as pattern use and recognition is already a large part of the primary maths curriculum. In some cases binary numbers were used to teach patterns in maths. There were also many explicit descriptions of students noticing and understanding patterns.

Students were observed learning about, or demonstrating this concept through recognising patterns in the binary numbers activities; recognising that a pattern was being used in the parity card trick; taking understanding gained from one type of problem or algorithm, and applying this in a new situation (e.g. from KidBots to Scratch, from binary numbers to base ten numbers); and recognising that they could take part of an algorithm and repeat it in another situation (both in programming and Unplugged programming). Some examples of the sections tagged as patterns and generalisation are:

- Describing students writing Beebot (i.e. turtle-style) instructions for drawing shapes of numbers on the ground, or moving a counter from a start to an end point — “there were some kids that could see things like if they could do a six they could just flip it upside down to do at nine as well. Actually, being able to see that they can tweak their program in small ways and it would do something else that they were aiming to do as well.’

- Describing what students had learnt or improved with — “Things like patterning, sequence and order.”
- Describing concepts students understood well — “noticing the doubling of each binary number and starting to get a feel for the patterns as consecutive numbers were created.”
- Describing students using Scratch — “It’s really cool that they have learnt through one experience how they can adapt that to do other things.”
- From feedback, concepts understood well — “Doubling of binary, patterns in creating consecutive counting numbers”, “Doubling of numbers to make the next binary number. Understanding that the maximum number that would be made with binary was double the largest number you had minus one, eg If 16 is the number. Double 16 minus 1 gives 31. Starting to see the patterning developing with the flipping of the cards.”
- From feedback, a concept that was understood well and applied to three different activities — “The concept of eliminating possible scenarios repeatedly by half”.
- From feedback on programming, concept understood well — “using a loop to make sequencing easier”, “Nested loops and parameters.”

One of the observations which stood out was made by a teacher with a very young group of students. These students were in year 1 and 2 (5-7 years old), and their teacher was excited and surprised by the results of teaching them binary, particularly for the children in their class who struggled with maths. They described students using patterns and generalisation to great effect, and benefits this had for their numeracy skills:

“We used [CS Unplugged] as a follow on from our numeracy. We went into look at the binary numbers first, I started with that, and that gave me some quite surprising results. Some of the children that struggle with numeracy picked up the patterns in the binary numbers really easily. So I had some of the lower ability children teaching some of the higher ability children because the higher ability ones just couldn’t get it.

That surprised me... They picked up patterns for other things. By looking at the way we did the binary numbers to start with, and the addition, they then started looking at other numbers and finding the patterns that came out... It was good for their addition and counting upwards. Helping them to count from 1-10, they had to think more about the actual pattern of the numbers...I was surprised because it showed them all the different patterns that addition and subtraction have if you look at them closely.”

Collaborating

There was an enormous focus on collaboration and teamwork among students in the interviews, and observations of collaboration occurred at least once in almost every different CS Unplugged and programming activity. Both the CS Unplugged activities and programming had either encouraged or required students to work together and collaborate. Students worked together when coming up with strategies to approach problems, and algorithms to solve them; designing and writing programs; working through programming resources (e.g. Code Club) together; taking on different roles in a team and sticking to these; teaching each other and sharing their learning; and working on their social skills through teamwork and learning from each other. Some examples of the sections tagged as collaborating are:

- From feedback, what students understood well in various CS Unplugged activities:
 - “Team work. Accepting a leader in the hoop activity [version of the Sorting Network]. Listening to others. Not just looking after yourself and your own position but to work towards the whole team achieving”
 - “The need to work cooperatively with others”
 - “I was amazed at how well the children worked together to solve the problems”
- From feedback, skills students identified as important and they had done well:
 - “Helping the ‘team’ to become competent not just having one person take over”
 - “communication, team work, considering others, helping each other, cooperating”
 - “communication by talking and using gestures, team work, thinking of others not just yourself (and your own position)”
- Describing students programming together
 - “In a lot of those activities they have to be working with other people so it is also really good for them to work on not only something that they love but also work on the social skills and the leadership skills”
 - “All those interpersonal dynamics all came as an opportunity to develop while they were creating something”

Some teachers reported that their students were already very good at working together so it was not these activities which caused this collaboration to occur, but in all of these cases teachers said the activities supported and strengthened these skills. In other cases, teachers described both specific students and entire classes who struggled greatly with working with other people, and in some cases strongly disliked working with others. These teachers

reported that these activities helped change that.

Two sets of feedback on a particular class gave an insight into their collaborative skills improving. The first set of feedback about this class stated students struggled with “Working well together as a team. A number just wanted to have their idea actioned rather than considering the opinions of others in the group even though they realised they did have to work as a team to get a team result”. The teacher wrote that “children in this class in particular would benefit from a number of similar activities to to work more coherently together in cooperative problem solving situations”. The feedback on a subsequent session with this class was very different. The teacher described how the students were “Working as a team and considering each other’s ideas.... interestingly I felt they listened more to each others ideas and considered their points of view and options better”.

Another teacher described how their students attitude towards teamwork had changed over the year:

“I mean the working together was a big thing, because mine can’t/ they could not work together. Then having to be taught how to work together and then having to work together to solve a problem... and it was something they enjoyed doing, to solve a problem. So, I really liked that. They’ve worked together in writing and create stories together now, which they wouldn’t have done at the start of the year because they didn’t like working together.”

One of the teachers worked with a target group of students, for one hour a week outside their usual classroom program. This teachers interview stood out as they described a situation where students developed and demonstrated excellent collaborative skills. This group of students had been identified by their usual classroom teachers as people who would likely benefit from some extension in this area. They were not selected because they were high achievers in certain subjects (although some were) and according to the teacher who worked with them they were a mixture of different personalities and abilities, and initially were not collaborative. However, as the students worked on projects in these classes this changed.

“[there were] the really charismatic ones where image is everything, so they came to these a bit guarded. Then you’ve got your real quirky, ‘nerdy’ kids as well. We had the shy girls, we had the dominant girls, I’m talking a real mix... and balance of boys and girls, it was across the board.”

“... what happened over time, we’ve got a real team spirit now. Next thing they’re starting to turn to each other to solve problems, so you’ve got the ‘cool kids’ and the ‘nerd kids’... They’re all valued in the group because at one stage or another they’ve actually all had something to offer the group, or someone is a bit of an expert”

“I think they valued collaboration, they valued that to produce something really good... by yourself you can reach good, together reach great, and I think that’s almost their mantra going forward.”

This group of students subsequently started their own Code Club so they could work together to teach other students.

Creating

Creating was the most common CT approach identified, through both students making things inside and outside of class, and teachers describing connections between the topics being taught and students creative thinking. Those students who worked with Scratch and ScratchJr, and were given relatively free reign with these, consistently used them for a creative approach. Both during and outside of school time students made animations, games, and artwork with these programming environments, and would enthusiastically share them with their teacher, friends, and family.

Students were observed learning about, or demonstrating this concept through designing and creating original digital artefacts, most frequently through programming; creating artwork based on CS concepts from the Unplugged activities; and using CS and programming concepts and resources when creating in other areas, such as story telling and class presentations. Some examples of the sections tagged as creating are:

- Describing students during class time:
 - “We did a lesson on binary. Then the exciting next step we did was turn it into the alphabet stuff and we made a piece of artwork of our names. We would visually represent the letters so it made a special shape.”
 - “We moved on from there too some of the activities that were around creating codes through binary for pictures or colours or sounds etc. They created them for their families as well and took them home for their families to decode”
 - “Any time they can get on the iPads they are creating things on ScratchJr”
 - “They’re great when it comes to inventing things... [they are] creators, inventors”
- Describing students creating outside class time:

- “they’re programming outside of school, they’re spending huge time gaps creating what they’re doing”
- “Lots of [parents] have commented and said that rather than their kids just playing on the computer at home they’re now going home and coding something and then playing it, so they’re a lot happier because they’re creating rather than just playing”
- “An example of how the kids embraced it is it when we had a pet day, it’s a day where we also have a kind of show type thing where people make things like... craft stuff, and one kid made a Scratch Jnr thing at his house and then he entered it in the show... he went away and had that on his own family iPad and was excited to make a little Scratch Jr movie, and it was really good. A huge amount of interest from him. It was an undersea type scene with fish... It was really cool because he had learnt it at school, gotten really excited about it, then gone home and organised what he needed to do in order to explore that in his own free time.”

Several interviews described how creative programming gave students a new way to express themselves and their learning. This was beneficial for many students, but some teachers noted that there were particular students in their classes that this had a profound impact on. These were generally students who struggled with more traditional methods of presenting their work, such as writing and speaking.

“It gave them a different way of learning, or a different way of expressing what they had learnt. We have a couple of kids in here who are... they just really struggle with writing and presenting things. Doing a poster to show their work, it is their worst nightmare. But if you ask them to create something on Scratch to present their findings then they do such an amazing job”

“I think one of the main things is how they can represent their learning in different ways, more challenging ways. And the creative side of it, you’re not just regurgitating something. As well as creative thinking it involves a lot of logical thinking, because some of them didn’t really have a lot of that, they were quite weak on that, and the creative side as well. I think that’s really important. That’s what the programming offers you too, a different way of representing your creativity and your learning.”

“[the] children who have English as their second language did some amazing story telling because the ScratchJr program could do the dialogue and speech bubbles for them, or they could go away and record the sounds and then put them in. They were making some very high end sort of stories, and these little girls who don’t say boo in class [meaning ‘don’t say anything’] were just very quietly creating this amazing presentation. I was just blown away”

Describing a student with severe learning difficulties: “the creativity that this kid had was amazing but the way in which she could express it using Scratch... and then her explaining what her story was, it was just cool... she created some amazing stuff”.

One of the teachers shared an experience from their class that they were particularly impressed with. A student used their experience with ScratchJr to create special effects for a play that students were writing and performing in class:

“We started a play based program last year and one of the children had made a little hospital set up... someone went away and made a ScratchJr thing that made a little ticking noise like a heart monitor. They went away and designed this thing that kept ticking, I don’t know how they did it. So, they incorporated these little repetitive elements into the ScratchJr program.

They made this little thing and said look it’s the heartbeat, and I just thought oh my goodness, I never would have thought of that in a million years! It was really cool. They are now integrating their knowledge of that code and integrating it into their play. It was done seamlessly and naturally. It reinforced the play that they were trying to engage with but also just supported it, rather than replacing it. It was their knowledge, they had it and they used it to make their play better. I never would have thought of it and I think that’s kind of cool for them to be able to have that freedom to use that.

In terms of using Scratch, that could be a format for them to share learning from other contexts and if we free them up and say that they can share their knowledge of something they’ve learnt about another context. If they are able to share through that mode, who knows what they will do. Because if they’ve got the knowledge it will just be another tool in their backpack that they can go to. They can say well I can

use Scratch for that.”

Debugging

Debugging was often tagged along side Persevering, as having to debug was something students were challenged by and which they frequently worked at for long periods of time, even when they became frustrated with the process. Students were observed learning about, or demonstrating this approach through programming (original programs and from resources such as Code Club projects); creating algorithms; and when carrying out algorithms or instructions they were given to follow (such as KidBots and the Sorting network). Some examples of the sections tagged as Debugging are:

- Describing students using the Bee-Bots- “During these sessions some children were beginning to understand that what they thought they had programmed was different sometimes, and they were able to then sort out the problem through either working by themselves or together”
- Describing students doing the KidBots activity —“Some of them would write their program and give it to their partner to do, and then it wouldn’t work out so they’d have to figure out did this person read it incorrectly or is your program incorrect”, “The concept of debugging was also a strength for this lesson”
- Describing students becoming more persistent with debugging —“they were starting to talk amongst themselves about what to do, ‘What haven’t we done?’, ‘Why is it doing that?’. Initially they were just saying ‘Oh it doesn’t work’, expecting you too fix it. After that they started to actually look at what was going on, what they were doing, and that what they were doing was making it do this. So if I’m telling it to do this it’s going to do that, but if I then tell it to do something else it does something else”
- Describing students programming —“One student is a really talented programmer, he’s the one that can debug your program and solve that”
- When doing the Sorting Network activity —“when it didn’t work, to figure out what had happened, we would backtrack and they would find that someone should have gone one way but they went the other way”, “You’d always see children who thought ‘oh I’ve got the biggest number’ so therefore I’ll go there, and it would kind of muck it up so you’d just let them go and at the end they’d think ‘hmmm... these aren’t

exactly in order, what's happened?' ”

A common theme which came up in sections about Debugging (and also Persevering) was teachers discussing how when students couldn't figure out what was going wrong with their program, they would ask for help and expect the teacher to do it for them, only to be shocked when their teacher told them they couldn't help because they didn't know the answer either. The teachers who described this generally did not see this as a negative, and as time went on they found that the majority of their students began to work more independently on solving their problems. Teachers would also debug alongside their students, or as a team with the class, for example:

“... they'd ask 'how do you do that?', 'I don't know! Remember I'm learning too!' So, I would have it up on the board and I'd be trying it at the same time as them. Then they would be watching and one would put their hand up and say 'wait you need to put that one in there'. It was quite cool doing that, and then I just stepped away and let them carry on. So they're probably more efficient than me now”

When saying “I don't know!” the teacher had a happy and excited tone of voice, and was portraying this as a positive thing.

Persevering

While Persevering was the approach tagged least often , and the second least tagged skill overall, it was in fact a strong theme, and the number of times it was tagged does not truly reflect how much teachers emphasised it. They described students becoming more determined, more likely to take charge of their own learning, and less over-reliant on them as their teacher. The observations of persistence were particularly intriguing, as teachers believed this was directly related to these lessons, and some of their students carried this across into their work in other subjects. There were also teachers who stated their students were already generally persistent before they started teaching these lessons, but they also said that the lessons supported their existing skills.

Students were observed learning about, or demonstrating this concept through carrying out algorithms and creating them, particularly when working with directional instructions with Bee-Bots and Kidbots; and when creating original programs, or programming from resources (such as Code Club projects). Some examples of the sections tagged as Persevering are:

- Describing how students changed over the course of the year — “they became a lot more autonomous, and they could choose what to do, so I didn’t have them constantly coming up to me with a problem, which you normally have. ‘I can’t do it, I’m stuck, what do I do now? Provide me with the answers’. [Now] a lot of them would say ‘I can’t do it’ and they would go to their buddy and they would say ‘well have you tried doing it this way?’ It was quite nice to have them not be so... well to have them become more self-reliant. Especially at this age. They’re a lot more self-managed and a lot more cooperative with each other, which is a big thing.”
- Describing students programming in class — “They found that hard. But they persevered and we got there”, “if a kid came up to you and said ‘ohh how do you do this?’, if you just threw it back to them and said ‘why don’t you figure it out and come back and teach us?’, [then] the child would”
- Describing benefits to students — “Rather than giving up on stuff, trying to work at problems and trying to figure out the best solutions. That’s where this fits in quite nicely because of the computational [sic] thinking, and it’s fun as well”
- One class had a Scratch programming book, and one of the students wanted to work out his program without using it — “I watched one of them and he didn’t want to do the angles [tutorial], he wanted to work it out himself but then it kept going the wrong way... but he just kept trying and trying, and trying to figure out the x and the y. But it wasn’t working so in the end he had to succumb and go check the book and find out what was wrong. But he was really persistent.”

Several teachers discussed how some of their students were not persistent in other subjects, and had a tendency to give up when they found things challenging, behaved very differently in these classes. They worked much harder, and continued working even when they encountered challenges. Teachers were particularly happy about this, and they felt proud of these students. A teacher described how in their class one student in particular was struggling with programming. They moved this student back to work with the Bee-Bots, and when this was also highly challenging they tried giving the student the Beebot iPad app to try instead:

Teacher: “Some were progressing quite quickly, and I would work with the ones that were not because they were struggling. For those ones, we ended up leaving the coding and used the Bee-Bots instead. Also one of them still really struggled with it so I got the Beebot pyramid app... she did it with a counter that we put little eyes on. She would move the counter first before she did the actual programming on the iPad, so that she had an idea of what she was doing”

Interviewer: “Would you say that they [referring to the class] were successful or unsuccessful?”

Teacher: “I would say they [referring to the class] were successful... and while it frustrated Amy³, who was the one that worked with the Beebot [app], she didn’t give up, which is unusual for her because when it’s too hard she would normally give up, but she was determined to get it to the end... Normally she would go ‘too hard! Not doing it!’ and she would go off in the corner and cry. But she was like... she was really good with it.”

Tinkering

Tinkering was frequently tagged alongside Logic and Creating, as these three all went hand in hand with exploration and play. Teachers described students being excited by their learning, and how because of this they would repeat activities, want to find out more about them, and would try different ways to do them. They allowed them the freedom to modify activities and extend them, and found that students created and solved complex problems by doing this.

Students were observed learning about, or demonstrating Tinkering through exploring programming environments; repeating and expanding on the CS Unplugged activities; building on projects they had done at school in their own time; using trial and error approaches to their learning and in activities; and through creating digital and physical artefacts that illustrated their learning. Some examples of the sections tagged as Tinkering are:

- Describing students ‘playing’ with programming in class:
 - “They just played a little at first and then trialled some of the things which was really good, and some of them have actually carried it on into their [other] classes”
 - “They seem to have understood it a lot faster than I expected them to... They just play and work things out, they will come up to me and say look I worked

³ Name has been changed

out how to do this”

- “Any time they can get on the iPads they are creating things on ScratchJr. We haven’t taught them how to use that, we have just shown them program. They’ve just learnt through other children and helping each other”
- “they’ve had the Bee-Bots in their classrooms. So every morning when they come in it’s there for them to engage with. The children in turn, because parents often drop their children off, the children will show their parents or their grandparents”
- “I found that I could just give them a little bit of information and they would just go ‘oh I can do that’, and be off and say ‘come look at this, come and see this!’, and they will have done something else, they will now be making two wheels turn or something. They just went off on those and it was good, quite cool to see.”
- “The kids just had a play and found out a lot of the stuff that they could do with that.”
- Describing students doing Unplugged activities:
 - “The other activity we did was the parity trick, they love that, that’s on my whiteboard and the kids use it all the time. So it’s just there and I see kids doing it with other kids all the time, any parent that comes and they do it to them, they call it a magic trick.”
 - The barcode trick: “the kids were blown away by it and keen to do it again. They tried it with all different barcodes around the room.”
 - “They loved writing their name in Binary and also went on to create a secret message for class mates and some even went home and made a secret message for their parents!”
- From feedback, concepts understood well by students —“Not be afraid to test their program”, “a bit more risk taking started to happen. Not being so scared to do something and see what happens.”

Many teachers encouraged their students to play and experiment when they were learning about programming, rather than always using specific lesson plans. They described how their students would learn by trial and error, and through teaching and observing each other. When learning about binary and data representation, students would frequently write binary messages for each other, and create different ways of representing binary digits, such as with sound and lights. The parity trick and barcode trick were frequently repeated by students, both to show these to other people, and to test if these tricks would always work

or if they could ‘break’ them. Several teachers also stated that their students had become more likely to take risks, and were less afraid to test things when they were not sure what they would do.

8.3.5 *The Key Competencies*

As discussed in Chapter 4, section 4.2.2, a core component of the NZ curriculum is the set of *Key Competencies*, which students are supported to develop throughout their school education. The five competencies are: Thinking; Relating to others; Using language, symbols, and texts; Managing self, and; Participating and communicating. The inclusion of, and emphasis which is put on these Key Competencies is relatively unique to the NZ curriculum, and during my work many teachers remarked on what an important and beneficial part of the curriculum they are. The Key Competencies were not something I had considered investigating before these studies began. During this wider teacher study it became clear that there was a strong connection between these and the CS and programming lessons. Several CT skills correspond directly with the descriptions of the Key competencies in the New Zealand Curriculum:

- **Thinking: Logic, Evaluation, Debugging, and Tinkering**

“Thinking is about using creative, critical (*Logic, Evaluation, Debugging*), and metacognitive (*Logic, Evaluation*) processes to make sense of information, experiences, and ideas.”

“These processes can be applied to purposes such as developing understanding, making decisions (*Evaluation*), shaping actions, or constructing knowledge (*Logic, Tinkering*). Intellectual curiosity is at the heart of this competency (*Tinkering*).”

“Students who are competent thinkers and problem solvers actively seek, use, and create knowledge (*Logic, Tinkering*). They reflect on their own learning, draw on personal knowledge and intuitions, ask questions, and challenge the basis of assumptions and perceptions (*Evaluation, Logic, Tinkering*).”

- **Relating to others: Collaboration**

“Relating to others is about interacting effectively with a diverse range of people in a variety of contexts. This competency includes the ability to listen actively, recognise different points of view, negotiate, and share ideas (*All Collaboration*).”

“Students who relate well to others are open to new learning and able to take different roles in different situations (*All Collaboration*).”

“By working effectively together, they can come up with new approaches, ideas, and ways of thinking (*All Collaboration*).”

- **Using language, symbols, and texts: Abstraction, Algorithms, Creating, and *programming***

“Using language, symbols, and texts is about working with and making meaning of the codes in which knowledge is expressed. Languages and symbols are systems for representing information (*Abstraction*).”

“People use languages and symbols to produce texts of all kinds: written, oral/aural, and visual; informative and imaginative; mathematical, scientific, and technological”. (*Many related to one or more of: Abstraction, Algorithms, Creating, and programming*)

“They recognise how choices of language, symbol, or text affect people’s understanding (*Abstraction*) and the ways in which they respond to communications.”

“They confidently use ICT... to access and provide information and to communicate with others”. (*While this competency is related to the use of ICT rather than understanding of it, it has a connection to Creating, and ties in well with teachers’ observations of Creating*)

- **Managing Self: Persevering, Collaboration**

“This competency is associated with self-motivation (*Persevering*), [and] a “can-do” attitude (*Persevering*)”.

“Students who manage themselves are enterprising, resourceful, reliable, and resilient (*Persevering*).”

“They know when to lead, when to follow (*Collaboration*), and when and how to act independently. (*Persevering*)”

- **Participating and Contributing: Collaboration**

“This competency is about being actively involved in communities... This competency includes a capacity to contribute appropriately as a group member, to make

connections with others, and to create opportunities for others in the group” (*All Collaboration*)

Though it is not commonly listed as a CT skill, programming is also included in this list, as there is a clear connection between this and the Competency of “Using language, symbols, and texts”. This same competency also had connections with the CT approach of Creating, which is frequently demonstrated by students via programming. As explored in section 8.3.4, creating via programming provided students with a new way to present and communicate their knowledge, which is a key part of “Using language, symbols, and texts”.

The links between these topics and the Key Competencies were explicitly referenced by several teachers, in both the feedback and interviews:

- “Fantastic activity, can’t wait to do more. I really see the benefit of these. They are working on Key comps/values as well as comp sci” — Feedback entry
- “I think more than anything, this is highly dispositional. I refer to the key competencies with them. Because you’ve got your using ‘language, symbols, and text’. You’ve got your ‘participating and contributing’. It’s a very social activity as well, as far as the kids relating to each other. The deeper thinking, creative thinking. The critical thinking is a big one too. Because we talk about bringing the Key Competencies to life, the NZ curriculum states that it’s highly important, but then that’s where teachers become stuck sometimes, with how do you explicitly teach or encourage those key competencies? But this has been amazing” — Interview with a teacher
- “The key competency of thinking... we think we give students a lot of opportunities to develop that thinking, but I don’t think we do it as explicitly as the computational thinking side of stuff does” — Interview with a teacher

One of the teachers who discussed the Key Competencies at length described that the way these new subjects supported the Key Competencies meant they would be easier to integrate into the curriculum. They said this could also help convince other teachers and principles that it was worth dedicating time to these subjects. When asked what some of the challenges they faced during this time were they responded:

“I guess the challenges would be knowing that I’m ultimately going to have to justify why we’re doing this. The challenge has been making the connections in order to justify to colleagues that this is worth them spending their time on. And I know it is, but it’s how to frame that to them... that’s why focusing on the key competencies has been my way of selling it to the other staff, and they’ve seen a change in the kids, that in itself is something.”

The cohesion between the Key Competencies, and the Computational Thinking Concepts and Approaches, made NZ schools a unique setting for integrating these new topics, CS and programming, into an existing primary curriculum.

8.3.6 Problem Solving

Unlike the Key Competencies, problem solving was a topic I had intended to investigate from the beginning of the study. In the interviews teachers were asked specifically whether they had observed impacts on students problem solving skills. Of these 18 teachers, 11 said they believed their students’ problem solving skills had indeed been improved by the programming and CS Unplugged lessons. Two teachers said they were not certain, but they thought the lessons probably did have some benefit for problem solving skills. The remaining five said they could not give any answer to this question. They felt there was no way they could be sure whether the lessons had, or had not, impacted their students problem solving skills. They said this was because it was impossible to separate the potential impacts (specifically on problem solving skills) of these lessons from every other subject students had worked on in class. All teachers agreed that the lessons did not have a negative impact on problem solving skills.

Additionally to this specific question, problem solving was referenced in several interviews and entries in the feedback form. These were often linked to descriptions of cross curricula learning and learning transfer, which are discussed in the following section. Discussions on problem solving skills also frequently connected with the Key Competencies. This makes sense, as the competency ‘Thinking’ expressly covers general problems solving skills, and the others each refer to using the competency to do things like: confront challenges, work towards goals, and use their knowledge to create solutions. Some examples of sections tagged as problem solving are:

- “They are also asking good questions and thinking creatively in how to problem solve the task” — From feedback
- When discussing benefits to students in interviews:
 - “The problem solving! I quite liked the problem solving because you can give them quite complicated things”
 - “It’s very good for their sequential thinking and their problem solving”
 - “I think it’s the problem solving skills, when you look at the logic in it, and the ability to actually see a pattern, or a path, an outcome and those steps to get to it... I think it is that problem solving and that reasoning”
 - “I think because it’s all part of the problem solving, and they’re doing problem solving in their maths and other areas so it all fits in quite nicely with the curriculum and everything we’re doing in the classroom... again it was that problem solving, and having success.”
- When asked in interviews if their students’ problem solving skills were impacted:
 - “Yes, it encourages them to think in different ways... they definitely had to use some trial and error for things like the binary activity. Same with that muddy city, that trial and error, give it a go, work out what else you could try.”
 - “Initially no, but after we’d done it for a few weeks they were starting to talk amongst themselves about what to do, ‘What haven’t we done?’, ‘Why is it doing that?’. Initially they were just saying “Oh it doesn’t work”, expecting you too fix it. After that they started to actually look at what was going on, what they were doing”
 - “...they’re skills that they can be using across a range of things like problem solving, critical thinking, being creative”

8.3.7 *Cross-curricular learning*

There was a large amount of cross-curricula integration in this study, and a small amount of learning transfer. Teachers integrated the CS Unplugged and programming lessons with a wide range of other subjects. These subjects were maths (numeracy, patterns, and geometry), literacy (reading, writing, oral language), science, languages, geography, physical education, art, and design. Teachers with a particular focus on encouraging the Māori and

Pacifica students they worked with described how they integrated language and cultural context into their lessons. Some examples of this cross-subject integration are:

- “I ended up teaching **geometry** in Scratch... They had to learn how to do the angles, so they had to work out the angles of the inside and outside of the shapes in order to get Scratch to do them... It was a really good way of teaching **geometry**”
- “[there is] the obvious relationship with **maths** and **technology**, but there’s also the fractured fairy tales and the other things with **literacy**, there are a lot of things you can integrate it with”
- Describing students’ Science Fair projects — “Then you’ve got another kid who actually designed a lab and you press different buttons and the nails go in the coke, and sometimes one will go rusty and the other will go ... whatever. Or you’ve got kids who did mouldy food, that kind of thing. What was quite encouraging was anecdotal evidence that they [their usual teachers] had noticed an improvement in the quality of their work in **science**, because it’s the reinforcement. That’s like, wow that’s awesome, let’s do more of that!”
- “... they would do all the directional type things. It’s so good for the **oral language** skills, the instructors had to have really good oral language.”
- “When I came back I wanted to do this **Te Reo Māori** ... I wanted to see how successful I could be in using programming to teach a language. So I had a project going on where you’d show an image of something, like a well-known building, like the Sky Tower in Auckland, and then in Māori you would ask where that was. We had our character going all over NZ... [a lot] of our students could not have marked on the map where Auckland was. So, we were teaching that Tāmaki-makau-rau is the Māori name for Auckland, but we were also teaching ‘where is that place?’ ”
- “I could almost say that I could go into the classroom and just use Scratch to teach ... I could just use it all day!... I feel like I’ve taught **Science**, **Geography**, you’re teaching **literacy** all the time, you’re teaching **numeracy**, **Te Reo Māori**... There’s a lot of learning going on in this room.”

- “We used [CS] as a follow on from our numeracy. We went into look at the Binary numbers first... By looking at the way we did the binary numbers to start with, and the **addition**, they then started looking at other numbers and finding the patterns that came out... It was good for their **addition** and counting upwards. Helping them to count from 1-10”

After her three years of teaching CS and programming, Angela described how she felt about teaching with Scratch: “I almost think ‘what couldn’t I teach’, let alone to see all the thinking and the shift in their thinking”. Teachers reported that the CS and programming supported learning in these other subjects, gave students a new way to process information, and new ways to present their learning. Not all teachers reported these same levels of cross-curricular success, but none said there were negative impacts on either subjects. The challenges of time constraints on class time (and preparation time for teachers), or the already very full primary school curriculum were brought up by every teacher who was interviewed. However, they said that integrating CS and programming with other subjects made fitting this in to the curriculum more doable.

These topics were also frequently integrated into students ‘inquiry’ topic, in classes that took an inquiry based approach to learning.⁴ One teacher discussed the links between these subjects and their students inquiry topic, which was based on the Christchurch rebuild. In this they investigated the ‘gap-fillers’⁵ in Christchurch City Central, and then created their own gap-fillers around their school. During this inquiry work students spent several lessons on the CS Unplugged Muddy city problem, which their teacher then linked to the Christchurch rebuild:

From feedback form —“Great linkage with their unit on the Christchurch rebuild regarding networks, even though their focus [for this unit] is on the Arts. [They] could create walking networks which took you through the different gap fillers etc in CHCH. Good opportunity for them to work in groups and to build on group skills.”

In these lessons they expanded on the idea of city planning, building cardboard buildings and ensuring “that there was plenty of green space left for flowers and plants. They

⁴ https://en.wikipedia.org/wiki/Inquiry-based_learning

⁵ An ‘urban regeneration social enterprise in Christchurch’, where small installations are created around the city, where many buildings have been demolished post earthquake <https://gapfiller.org.nz/>

really wanted to do more with the Muddy City other than just put pavers down”. Scratch programming was then introduced to this topic:

“We’ve got all sorts of gap fillers around the school now. We’ve got geo-cache boxes... But what the code club did was choose 4 or 5 of the gap fillers they visited, then researched and found a photo and some information about them, then they coded a maze on Scratch where they could go around and find out the information about them, and that was based on one of the code club projects.”

“It was also great that we were able to select something directly related to their Inquiry. They have already made part of a Sound Garden for the school as a follow up to their visit to the city.”

Along with subject specific integrations, many teachers described how skills from these lessons were “cross curricula strategies” and transferred to other areas of students learning.

- Discussing the CS Unplugged activities —“Those are **cross curricular** strategies. They are fundamental strategies and are also life strategies. They make connections really easily across the curriculum”
- Discussing programming —“I think the main thing is that **cross curricula** approach... They want to show their learning, put a display up on the board and talk about what they’ve been doing... For me that’s the real measure of success, that they’re applying their learning across curricula contexts and extra curricula contexts”

8.3.8 *Learning transfer*

As discussed in Chapter 3, one of the main claimed benefits of CT is that the skills students acquire through learning CS, programming, and CT will benefit them in many areas of school and general life. It is seen as something which students will transfer to other subjects, and use it in their approach to general problem solving and learning. This is a claim that has not been thoroughly researched, and so was a topic I investigated in this study.

Eight of the eighteen interviewed teachers believed their students had transferred some of the skills they learnt through CS and programming, to other subject areas. Similarly to problem solving, some teachers had no comment on this, as they felt they needed to observe their students over a longer period of time or test their skills somehow. Others described benefits to students general learning, but only in the context of students social development

or cross-curricula activities, rather than applying skills in different areas. However, those teachers who did discuss it gave several interesting examples:

- **On transfer to maths:**

Teacher: “[For coding] I would say ‘go out and draw this problem out there’ and they could act it out, and they did it in maths as well with a problem. Because they said ‘I can’t work it out, can we go draw it out there like we do with code?’ and I said yes and off they would go. So, they were using it in maths as well.”

Interviewer: “Did they make that connection themselves?”

Teacher: “Yes... It progressed their maths a lot”

- **On general transfer:**

- “The kids could make easy connections to other areas and I think that shows how valuable it is... I can definitely see benefits and how they have learnt those skills and can transfer them.”
- “When they realised that the skill they were focusing on wasn’t computing as such, it was the thinking side of things, it really paid off for the kids that could do that because they did take it through into those other subjects”
- “Although [these activities were] just in one programme that can be transferred. Generally, kids today [aren’t very good] at transferring skills... it has to be explicitly taught. Just watching today, when they were doing that thing/ how I taught them that if they were going to step it through yourself then what would you do... I think it is slowly transferring, which is really good”
- “It is all about sequencing, working in a team, making sure you go step by step, all of that stuff that you need for everyday life... It’s not just coding it’s what we should be doing with everything. So hugely beneficial.”

8.3.9 *Student impacts*

Across all feedback responses and interviews conducted, teachers said the vast majority of their students were highly engaged with, and excited by programming and CS Unplugged activities. They described students who “loved” the activities, looked forward to the classes, and one teacher said that if they had to cancel or move a lesson on CS or programming then they would be “accosted in the playground!”. Many said that this was something relatively

unique to these subjects, for example one teacher said “the reason I kept going [with the activities] was I noticed how engaged the kids were. Even though they struggled at times to keep focus, struggled with that self-motivation, for some reason there is something in these activities that sparked them”.

The game-like, physical movement, and tactile nature of the CS Unplugged activities was frequently cited as a reason for this engagement. Teachers were asked if they felt this engagement was related only to the style of these activities, or if it was also related to the actual CS concepts. Some said it was due to both, while others were unsure. However, regardless of whether the concepts were related to this engagement, teachers reported that the majority of students had gained an understanding of the underlying concepts, and only a small number seemed to think of these as simply games. Several teachers described individual students in their classes who did not enjoy the content, or became bored of it, but in these cases they stated that these students were mainly disengaged as they had high learning needs.

Along with excitement, interest, and engagement, there were a set of similar observations of students that stood out. These were about the strong impact these lessons had on specific students, particularly on the students ‘you wouldn’t expect’. Teachers commented that students who were normally disengaged with learning became re-engaged by these lessons, developed leadership skills and confidence, and in many cases achieved far beyond their teacher’s expectations. These students tended to be described in one of several ways: quiet and less confident; ‘less able’; or disengaged with learning in general because they struggled with usual teaching and learning methods, particularly reading and writing. There were also several cases where teachers observed that students in their class with dyslexia were particularly successful. Teachers observed how some of these students embraced these subjects, and were empowered by them:

- “This task brought out the best of those that normally don’t share - I saw a side to children that I rarely see’.
- “Sometimes those quieter ones, they surprised me really just how well they engaged and participated in things... seeing the children achieving and those ones that aren’t always that fabulous at something, this was their thing, and they shone.”
- “I loved hearing from those children who would normally sit back and watch. This really engaged them.”
- “It was awesome to see some of the more learning challenged children so excited about their success and enthusiasm”

- “It was particularly pleasing to see some of the ‘quieter’ children having a go and showing they had some really effective strategies.”
- “It was interesting to note how the less able children concentrated hard on the task and were quite methodical in their approach, whereby some of the very able children tended to rush into it and make errors.”
- “One group in particular just whipped through both of those [Code Club projects] and were like ‘We’re done!’ ..and they were the group that you wouldn’t expect either... None were natural leaders in the class. Not really techy... For them the coding worked.”
- When asked about benefits to students — “It maybe tweaked some of the attitudes to learning for the kids who were a little bit switched off. Maybe it improved the motivation.”
- “I think it is tapping into those children for whom that is their thing... giving them a different way of thinking... You could just see the way that they enjoyed that”

For those students who usually were disengaged and found learning difficult, and in some cases had behavioural issues as a result of this, these lessons seemed to appeal to their way of thinking in a way that the normal curriculum did not. Some teachers who observed this commented that this may have been due to the novelty of the lessons, but regardless these students were successful with this content. One teacher described how a student in their class, who had previous experience with Scratch, became a leader in their classroom, and how much they changed throughout the year:

“One of the kids got that Scratch book for Christmas, so when I had the school version of it with me he said ‘I’ve done this and I tried it at home!’ So, he became my expert”

“... The child that became the expert was probably really quite immature for his age, [but] he’s now left my class this year as the leader... his mana⁶ just went up.”

“...that kid wasn’t really a confident kid... he became confident in quite a few areas.”

This story was not unique to this teacher and observations of students going through a

⁶ Mana is a Māori word with a range of complex meanings. In this context the teacher is describing how the classes respect for this student, his status within the class, and his self respect increased.

similar process were observed in several classes.

In contrast to these stories of unexpected students being so successful, many teachers said some of the least successful students in their classes were the ‘high achievers’, or ‘gifted and talented students’ who they thought would have thrived in this area. Many teachers described how this group of students often became frustrated with these activities and would give up more quickly than their peers. Others said these students approached the subjects with a slight degree of arrogance, rushed through the learning, did not pay attention to detail in their work, and at the end had gained less knowledge or achieved less than their classmates. Three submissions in the feedback form were from a teacher who ran several CS Unplugged activities with a ‘gifted and talented’ class.⁷ These were notable as this teacher had done these activities with a range of different classes and students before, and this group did not progress through as much content as other classes (of the same age group) the teacher had worked with. The students were highly engaged, but the teacher remarked that they tried to rush through the activities, missed steps in the process, and did not appear to gain as deep an understanding of the underlying concepts.

One interview in particular demonstrated this stark difference between these students when they worked through these lessons. They described their students’ approach to one of the CS Unplugged activities:

“what was really interesting to me... I have five profoundly gifted children in my class. They sat back and normally they do everything. [Normally] they know everything and everything is easy, and they sat back and these other children who I have never even seen this side of the them...

These are practical farming kids ... there was this group of five of them that basically just did it [the whole activity]. They are the dyslexic kids, the really problem solving kids... But it was fast and I am talking about kids - I have a boy who can barely read, very dyslexic, and he could just see it really clearly and a lot faster than I did when I did it myself.

That made me look at those children differently, and it was right at the start of the year and I hadn’t seen that side of them before. It was really interesting.”

⁷ These classes took place at Mindplus, a programme for gifted and talented students. <https://nzcge.co.nz/mindplus>

8.4 Conclusion

Throughout 2015 and 2016, I worked with a large number of teachers from a diverse range of school environments. These teachers were provided with training and resources for teaching CS Unplugged and programming, and shared their experiences of teaching this material with me. By collecting feedback on lessons and observations of students learning, I aimed to gather more information on how teachers can be supported to teach these topics, to evaluate the suitability of the resources used, and to investigate students development and use of Computational Thinking skills.

The first two of my research questions were: *Can Computer Science and basic skills be taught in a typical primary school environment?*, and *How suitable were the teaching resources chosen for use?*

Throughout this study teachers reported that the vast majority of their students were able to engage with the lessons, and did seem to be developing an understanding of Computer Science and programming. The reception of the resources and activities provided was highly positive, from both teachers and students. During this study, I found that teachers consistently, and successfully, integrated these topics into their teaching. They found that subject content could be made suitable for the different age groups, and students of all ages were able to engage with both the CS and programming (unplugged or ‘plugged’) content. Teachers were generally happy with the support they received. The main suggestions for how it could be improved were to have more focus on connecting teachers so that they could collaborate and support each other; and to build on current resources, or create more, which gave them a clearer idea of how their students should be progressing in this subject over time.

The third of my research questions for this study was: *Can students develop Computational Thinking skills through learning Computer Science and programming?* From my results, I concluded that yes, as a result of the CS Unplugged and programming lessons they took part in, students developed CT skills. All five of the CT approaches, and five of the six CT concepts were observed by the majority of teachers, in their classes. The one concept which was not observed frequently was *Abstraction*, which was likely due to the age group of these primary school students. There were however several examples of students using or learning Abstraction, indicating that although it was not as common as the other CT skills, Abstraction was still attainable for some students in these classes. Teachers reported that the use and development of these skills were directly related to the CS and programming

lessons, and in some cases was enforced by existing curriculum

My fourth question was: *Can learning Computer Science and programming, and developing Computational Thinking skills, have an impact on problem solving skills or general learning, for the majority of students?* One of the main positive impacts CT is claimed to have for learners is that it will improve their problem solving skills in both school subjects, and in their general life. This has been a contentious claim in the CS education community, as there was little research to support this claim. However, the results of my work lend support to this claim, as the majority of teachers believed that learning CS, programming, and developing CT skills had a positive effect on their students' problem solving skills. There were also observations of learning transfer occurring, including between subjects which had not been explicitly integrated with the CS and programming content. This was limited to a small number of teachers however, eight out of a total of 32. While these observations of learning transfer were not common, they may lend further support to the claims of CT skills positive impacts on students.

My final question was: *What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students?* Teachers reported that for the majority of students there were no negative impacts of these lessons, and the only negative impacts that had been observed was temporary frustration from some of the typically high-achieving students in classes. The CS Unplugged activities were described as highly engaging, beneficial for students teamwork and social skills, and fun. Programming in block-based languages, with Bee-Bots, and through the KidBots activity, was also highly engaging for students, and gave them a new way to show their creativity, present their learning, and to challenge themselves. Along with these general positive experiences, teachers identified individual students in their classes who became particularly engrossed with these subjects and experienced greater success than their teachers had expected. As a result of these positive impacts, all teachers who were interviewed stated they were planning on continuing to teach these topics in the future.

These reports from teachers provide evidence that the CT can have benefits for problem solving skills, and that CT skills can be developed by primary school students, through CS and programming education. As the results of this study were based entirely on reports from teachers however, no objective measurements of students CT and problem solving skills had been made. Because of this, I chose to conduct an intervention study in 2017, with the aim of quantifying students CT skills before and after they took part in CS and programming lessons.

Chapter IX

Intervention Study

After three years of working with teachers and students, I had found student engagement in Computer Science and programming lessons was consistently high. With adequate support teachers were able to successfully integrate Computer Science and programming into their classroom programme, and into other curriculum areas. Teachers frequently observed positive impacts on their students' learning, problem solving skills, and persistence in problem solving. Each CT skill (concepts and approaches) had been displayed by students, and teachers believed the CS and programming lessons had directly contributed to the development of these skills. These observations make compelling evidence that through learning CS and programming students can develop CT skills, and these skills can have a range of positive effects on their learning. This evidence is strengthened by the consistency of these observations across a range of classroom environments. However, changes in students' CT and problem solving skills had not been objectively measured during these studies.

This final study aimed to further investigate how programming and CS lessons impacted students' CT and problem solving skills, by attempting to measure the CT skills of students before, and after they took part in these lessons. Based on my previous findings, my hypothesis was that the lessons would improve students' CT skills, would improve some students' problem solving skills, and would not have a negative impact on students' overall learning or problem solving skills.

In this chapter I cover how this study was conducted, the types of data collected, and how this was analysed. I then report the results of the Computational Thinking assessment students completed, and the themes identified from the analysis of teacher interviews. I discuss the possible implications of these results, and the limitations of these. Finally, with reference to the research questions I aimed to address, I discuss my conclusions and the prior findings that these support.

9.1 Overview

Over 2015 and 2016 I focused on subjective data, examining teachers' opinions and observations. This provided valuable insight into the impact of these topics in classes from experienced professionals in the teaching field. For this study, I chose to again interview teachers in order to gain more information about students' classroom experiences, and to also quantitatively measure students' CT skills.

I worked with three schools during this study, which I will refer to as Primary School A, Full-primary School B, and Intermediate School C. Across these schools I worked with students from years 4 to 8 (ages 8-13). All of the teachers I worked with taught one class of students at their school. Primary School A was a decile 3 primary school, covering year 1 to 6. I worked with three teachers at this school, who taught year 4-6 (ages 8-11) students. Full-primary School B was a decile 9 full-primary, and so covered all year groups from year 1 to 8. I worked with four teachers at Primary School A. Two of these teachers taught year 4-6, and two taught year 7-8 (ages 11-13). Intermediate School C was a decile 7 intermediate, with students in years 7 and 8. I worked with six teachers at this school, four of whom taught year 7 (ages 11-12) students, and two who taught year 8 (ages 12-13) students. These schools were all state schools (i.e. not private) and greatly varied in ethnic make-up, and students' socio-economic backgrounds.

I trained these teachers in using Computer Science Unplugged to teach Computer Science and asked them to teach a set of these activities in their lessons. I also asked them to have their students use an online programming education platform to do programming lessons. Before they began teaching this content I had students from these teachers classes, and as many other classes at each school as logistically possible, complete the Bebras Computational Thinking Challenge as a pre-test of students CT skills. At the end of the school term, after all participating teachers had completed the CS and programming lessons, I had these students complete another Bebras Computational Thinking Challenge (with different questions) as a post-test of their CT skills. This study was limited to Year 4-8 students, as the Bebras challenges I was able to access were designed for years 4 and up, and were too advanced for younger students. I then interviewed the teachers who had participated and asked about their experiences with, and observations of the study. I also collected students' demographic information from their schools. The results of these tests and the interview transcripts were used to attempt to answer the following research questions.

9.1.1 Research questions

The goal of this study was to verify conclusions from the previous two studies by using an additional measurement approach, and working with a group of teachers who were not entirely self-selected.

My research questions were, therefore, similar to those of my previous studies:

1. Can learning Computer Science and programming develop, or improve, Computational Thinking skills for the majority of students? (RQ2)
2. Can Computer Science and programming lessons, and developing Computational Thinking skills have an impact on problem solving skills, for the majority of students? (RQ3.1 and 3.2)
3. Can Computer Science and programming lessons, and developing Computational Thinking skills have an impact on overall learning for the majority of students? (RQ3.3 and 3.4)
4. If these lessons do have an impact on students' Computational Thinking skills or their overall learning, do these impacts differ between students of different genders or ethnicities? (supports RQ2 and 3)
5. What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students? (RQ4)

9.2 Method

At each school, I worked with multiple teachers who each taught a different class. Teachers taught CS Unplugged and programming lessons throughout term 3 of the school year. These schools were chosen as they volunteered to take part, and they had not previously taught CS, programming, or CT material to their students. This applied to all classes in the school, not just those who were taught by participating teachers.

Before the school term began I visited each of the schools and conducted a one-day workshop with the participating teachers. During these, I covered the CS Unplugged lessons on Binary numbers, binary representation of text and images, the Parity card and Barcode tricks for teaching error detection and correction algorithms, unplugged programming with

KidBots, and the Sorting Network. Teachers were asked to teach the CS Unplugged activities on binary, error correction and detection, and unplugged programming. Binary image representation and the Sorting Network were given as optional activities, which teachers could use if they had time. I directed teachers to the additional data representation and algorithms activities from the “Classic version” of CS Unplugged, as at this point in time the new version of CS Unplugged did not include these. These additional activities were also optional extensions.

During these workshops I also had teachers register for teacher accounts on Code Avengers¹. Code Avengers is an NZ based Education Technology company, sponsored by the NZ Ministry of Education, who provide an online platform for learning programming targeted at students and teachers. I intended for teachers to use this for programming lessons, and discussed with them how I would like them to go about using it. Unfortunately however, during the study, multiple issues arose with this system and it was not useable for the majority of the classes. These issues are discussed in section 9.3.2. Teachers were asked to complete a minimum of ten lessons on these topics with their classes. A lesson was approximately 45-60 minutes. Originally, at least two of the ten lessons needed to be on programming, and it was recommended that at least three were. However, it was not possible for all classes to complete this many lessons on programming.

At the beginning of term 3, before the CS and programming classes began, a pre-test was conducted. Students participating in the study, and as many other classes as logistically possible, took this test at each school. The non-participating students were in the same year levels as the participating students and were used as a control group. At the end of term 3, a post-test was given to all students who completed the pre-test.

9.2.1 Conducting the pre and post-tests

The Bebras challenges from 2016 and 2017 were used for the pre and post-tests. The Bebras challenge is divided by age levels, and so year 4-6 students were given different versions of these tests, than those given to the year 7-8 students. Bebras gives different year group categories different names: year 4-6 are called ‘Castors’, and year 7-8 are called ‘Benjamins’. The four different versions of the tests will be referred to as:

- ‘Castors A’ — The 2016 version of the year 4-6 challenge.

¹ <https://www.codeavengers.com/>

- ‘Castors B’ — The 2017 version of the year 4-6 challenge.
- ‘Benjamins A’ — The 2016 version of the year 7-8 challenge.
- ‘Benjamins B’ — The 2017 version of the year 7-8 challenge.

During the pre-tests, students in each class were randomly assigned to either an A challenge (2016 version) or B challenge (2017 version), for their respective age group. In each class, half the students were given an A and half were given a B test. During the post-tests students who completed the A test as a pre-test were then given a B test, and vice versa. To indicate the order of tests a group of students did, the groups will be referred to as Castors A-B, Castors B-A, Benjamins A-B, and Benjamins B-A. For example, a student in the group Benjamins A-B would have taken the 2016 version of the Benjamins challenge for their pre-test, and then the 2017 Benjamins challenge for their post-test.

Some questions appeared in both the Castors and Benjamins tests, generally with a different difficulty rating, e.g. a medium Castors question may then appear as an easy Benjamins question. An example of the questions in the tests is the “Animation” question shown in Figure 9.1. This appeared in both the Castors and Benjamins A tests and is shown as an example question² for ages 11-12 on the Bebras website.³

The tests were carried out successfully at each school.

9.2.2 Additional Student Data Collection

In addition to the pre and post-test results, I collected the following data on students:

- Name (in order to distinguish between students).
- Gender, as recorded by their school.
- Year level.
- Class they would be in during the study.
- Ethnicity, as recorded by their school.
- Their most recent National Standards results in Reading and Mathematics, as recorded by the school.

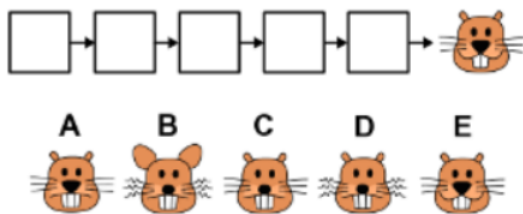
² As students were not given any information about the Bebras challenge, other than actually being given the tests to do, it is highly unlikely any participating students would have seen this question online before doing the tests.

³ <https://www.bebbras.org/?q=examples>

Animation

(Age group: Benjamins; Difficulty: easy)

B-taro is planning an animation, which shows a sequence of pictures of a face. The animation should run smoothly. Therefore, the order of the pictures is correct, if only one attribute of the face changes from one picture to the next. Unfortunately, the pictures got mixed up. Now B-taro must find the correct order again. Luckily, he knows which picture is last. He labels the five other pictures with letters A to E.



What is the correct order of the five other pictures?

- (1) $D \rightarrow B \rightarrow E \rightarrow C \rightarrow A$
- (2) $C \rightarrow B \rightarrow D \rightarrow A \rightarrow E$
- (3) $D \rightarrow B \rightarrow C \rightarrow E \rightarrow A$
- (4) $B \rightarrow D \rightarrow C \rightarrow A \rightarrow E$

Solution:

The correct order of the five other pictures is: (4) $B \rightarrow D \rightarrow C \rightarrow A \rightarrow E$

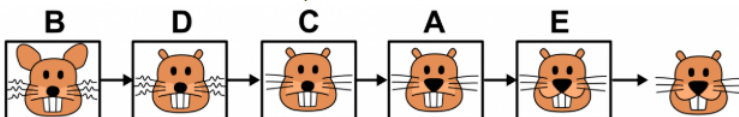


Figure 9.1: Example Bebras question ‘Animation’ — Reproduced from the Bebras International Challenge on Informatics and Computational Thinking website, www.bebbras.org.

- Students' scores in a pre-test and post-test.
- Information on their use of the Code Avengers website, including the amount of time they spent on each task, how many tasks they completed, and their scores in these tasks.

Students' Reading and Mathematics National Standards were collected to see if these had a significant impact on students' results in the Bebras challenge. These two measures were chosen as the Bebras challenge requires a significant amount of reading and reading comprehension, and past research has shown a generally positive correlation between maths skills and CS and programming ability.

9.2.3 Interviews

After the school term ended, each participating teacher was interviewed about their experiences and classroom observations. These were semi-structured interviews and were conducted in a similar way to the interviews described in Chapter 8. The one difference was these interviews were less open-ended than the previous, as teachers in this study had less freedom in choosing what they taught in class.

9.2.4 Data Analysis

Due to technical difficulties with the Code Avengers platform the data collected on students use of the website was unable to be used in any analysis.

Bebras Computational Thinking tests

For a full description of the Bebras marking process please refer to 6.4.1 of chapter 6.

After obtaining the results from the Bebras tests I first cleaned the data by following the steps:

1. Matched results to individual students using the names students entered on the test, and the class lists obtained from each school. This involved removing several entries where students had entered false names (e.g. "Cool Guy") or nicknames which could not be matched to full names.
2. Removed students who had not been given parental permission to participate.

3. Removed students who had only completed one of the tests.
4. Added the additional information of which school and class each student was from; whether a student was a participant or a control; their gender, ethnicity, Reading level, and Maths level; and which test group they were a part of out of Castors A-B, Castors B-A, Benjamins A-B, and Benjamins B-A.

I also removed one of the questions from the Castor B and Benjamins B tests, as it was deemed too ambiguous due to the way it was explained and displayed. This was based on mine and several of the participating teachers' judgement.

As previously described in section 6.4.1 of chapter 6, Bebras questions are rated as easy, medium, or hard, and the level of difficulty of a question dictates how many points are awarded, or deducted, for correct or incorrect answers to it. Each challenge is made up of five easy, five medium, and five hard questions. As assigning difficulty ratings to these questions has been found to be challenging and inaccurate in the past [23], I tested the accuracy of these original ratings. I took the number of times each question was answered correctly and ranked these. The questions were then assigned new difficulty ratings based on these rankings, with five questions of each difficulty. As previously mentioned one question had to be removed from both the 2017 Castors and Benjamins challenges, and so one category of difficulty only contained four questions for these two quizzes. The assigned difficulty ratings in the Castors A test were found to be accurate, when compared with the number of times they were correctly answered. However, all three of the other tests contained inaccurate difficulty ratings. The difficulty of a question needed to be changed five times for Benjamins A, seven times for Castors B, and eight times for Benjamins B. This reassigning of the difficulty levels is described more clearly by tables 9.2a through 9.2d.

The question "Hurlers Shake Hands" from the Benjamins B test was only answered correctly by 4% of students, a much lower proportion than any other question across all four tests, where the lowest success rates were 13%, 18%, and 24%. This question was therefore removed from the data set, leaving 13 questions in the Benjamins B test.

The Bebras marking system deducts points for incorrect answers. The goal of this is to discourage students from guessing answers. Students are all awarded a bonus 45 points, to avoid students getting a negative score if they answered all questions incorrectly. I chose to eliminate the bonus points and the point deductions, and give zero for incorrect answers instead. I did this as the majority of students attempted all, or all but one, of the questions

Castors 2016 Challenge			
Question	Original Assigned Difficulty	Correct answered	Final Assigned Difficulty
1. Birthday Balloons	Easy	56%	Easy
2. Bracelet	Easy	81%	Easy
3. Crane Operating	Easy	65%	Easy
4. Dream Dress	Easy	64%	Easy
5. Mushrooms	Easy	67%	Easy
6. Animation	Medium	44%	Medium
7. Biber - Hotel	Medium	45%	Medium
8. Geocaching	Medium	45%	Medium
9. Setting the Table	Medium	52%	Medium
10. Walnut Animals	Medium	46%	Medium
11. Beaver Gates	Hard	33%	Hard
12. Drawing Patterns	Hard	19%	Hard
13. Fair share	Hard	18%	Hard
14. Throw the Dice	Hard	22%	Hard
15. Trains	Hard	28%	Hard

(a) Assigned difficulties for the Castors (years 4-6) 2016 Challenge

Benjamins 2016 Challenge			
Question	Original Assigned Difficulty	Correctly answered	Final Assigned Difficulty
1. Bracelet	Easy	89%	Easy
2. Crane Operating	Easy	88%	Easy
3. Dream Dress	Easy	70%	Medium
4. Mushrooms	Easy	80%	Easy
5. Setting the Table	Easy	73%	Medium
6. Animation	Medium	43%	Hard
7. Binary Hotel	Medium	59%	Medium
8. Geocaching	Medium	66%	Medium
9. Trams	Medium	56%	Medium
10. Walnut Animals	Medium	77%	Easy
11. Beaver Dam	Hard	74%	Easy
12. Cross Country	Hard	30%	Hard
13. Drawing Patterns	Hard	42%	Hard
14. Fair Share	Hard	24%	Hard
15. Throw The Dice	Hard	43%	Hard

(b) Assigned difficulties for the Benjamins (years 7-8) 2016 Challenge

Castors 2017 Challenge			
Question	Original Assigned Difficulty	Correctly answered	Final Assigned Difficulty
1. Broken Window	Easy	91%	Easy
2. Falling Robot	Easy	29%	Medium
3. Ice Cream Machine	Easy	63%	Easy
4. Robot Exit	Easy	46%	Medium
5. Shelf Sort	Easy	56%	Easy
6. Bebras Painting	Medium	27%	Medium
7. Blossom	Medium	35%	Medium
8. Bottles	Medium	68%	Easy
9. Car Trip	Medium	22%	Hard
10. Tube System	Medium	13%	Hard
11. Beaver Code	Hard	48%	Medium
12. Mazes	Hard	22%	Hard
13. Party Guests	Hard	22%	Hard
14. Puzzle	Hard	78%	Easy

(c) Assigned difficulties for the Castors (years 4-6) 2017 Challenge

Benjamins 2017 Challenge			
Question	Original Assigned Difficulty	Correctly answered	Final Assigned Difficulty
1. Bottles	Easy	70%	Easy
2. Car Trip	Easy	43%	Medium
3. Puzzle	Easy	91%	Easy
4. Robot Exit	Easy	65%	Easy
5. Tube System	Easy	21%	Hard
6. Beaver Codes	Medium	77%	Easy
7. Mazes	Medium	18%	Hard
8. Paint it Black	Medium	12%	Hard
9. Party Guests	Medium	46%	Medium
10. Bus Stop	Hard	74%	Easy
11. Concurrent directions	Hard	42%	Medium
12. Hurlers Shake Hands	Hard	4%	Hard
13. Magic Potion	Hard	38%	Medium
14. Primary Health Care	Hard	25%	Hard

(d) Assigned difficulties for the Benjamins (years 7-8) 2017 Challenge

Figure 9.2: Reassigned difficulties for each test, based on how many students answered each question correctly.

in the tests so the possibility of losing points did not appear to impact their choice of whether or not to answer a question. Also, based on mine and teachers' observations it was not clear if students understood this point deduction system.

After reassigning the difficulty levels, removing the problematic questions, and changing negative marks to zero, all students' results were recalculated. Students' normalised learning gain was then calculated using Equation 9.1 from [91].

$$gain = \begin{cases} \frac{post-pre}{100\%-pre} & \text{if } post \geq pre \\ \frac{post-pre}{pre} & \text{if } post < pre \end{cases} \quad (9.1)$$

The tests used to investigate students' results and learning gain are described in section 9.4.

Interviews

As in the previous study, the interviews were transcribed and thematically analysed. The themes identified in these were similar to those discussed in Chapter 8, and observations of Computational Thinking were identified using the same method as in Chapter 8.

9.3 Schools and participant details

The total amount of useable data collected from each school is shown in Table 9.1. Data for students who did not return permission forms, and who only completed one out of the pre and post-tests, was discarded.

Two of the three teachers from Primary School A, all teachers from Full-primary School B, and five of the six teachers from Intermediate School C were interviewed. The other teachers from Primary School A and Intermediate School C were not available for an interview.

9.3.1 Schools and students

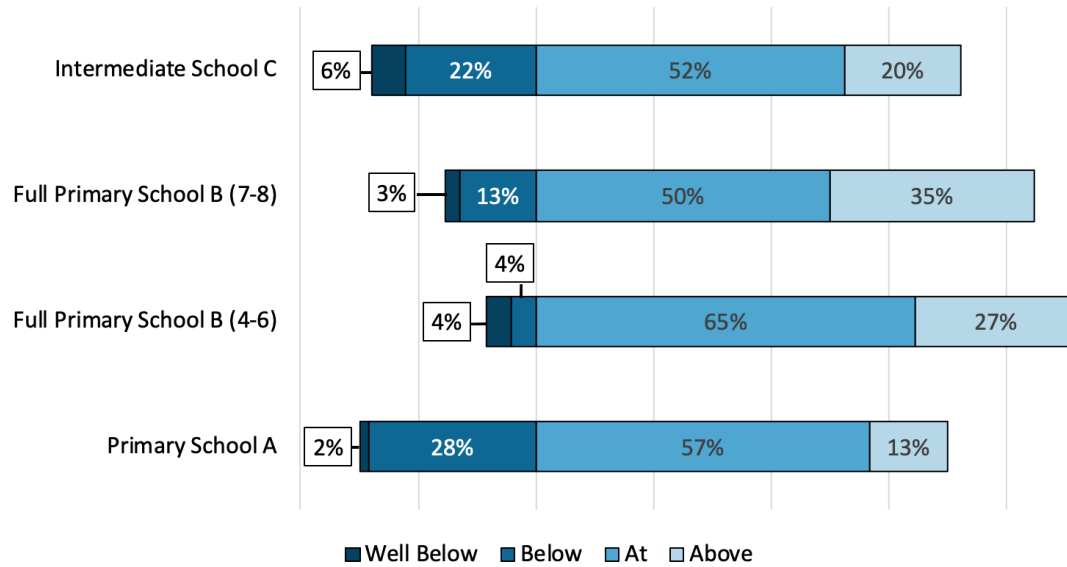
The proportion of female and male students within Full-primary School B (44% female, 56% male) and Intermediate School C (53% female, 47% male) was relatively even. At Primary School A, however, 63% of the students were female. This over-representation may have been due to not all classes in the school being included in the study, or may simply mirror the gender ratio within the school's total student population.

The proportion of students at each Reading and Maths level, at each school are shown in figure 9.3. This data was missing for a small number of students, so this data does not add to 100% for every school. The proportion of students at the Well Below and Below levels for Reading was much higher at Intermediate School C (28%) and Primary School A (30%)

Data type		Primary School A	Full-primary School B	Intermediate School C
		62 students	129 students	125 students
		3 participating teachers	4 participating teachers	6 participating teachers
Year	Valid	62	129	125
	Missing	0	0	0
Ethnicity	Valid	60	129	122
	Missing	2	0	3
Gender	Valid	60	129	122
	Missing	2	0	3
Reading level	Valid	60	126	122
	Missing	2	3	3
Maths level	Valid	60	122	122
	Missing	2	7	3

Table 9.1: Useable student data collected from each school

Reading levels at each school



Maths levels at each school

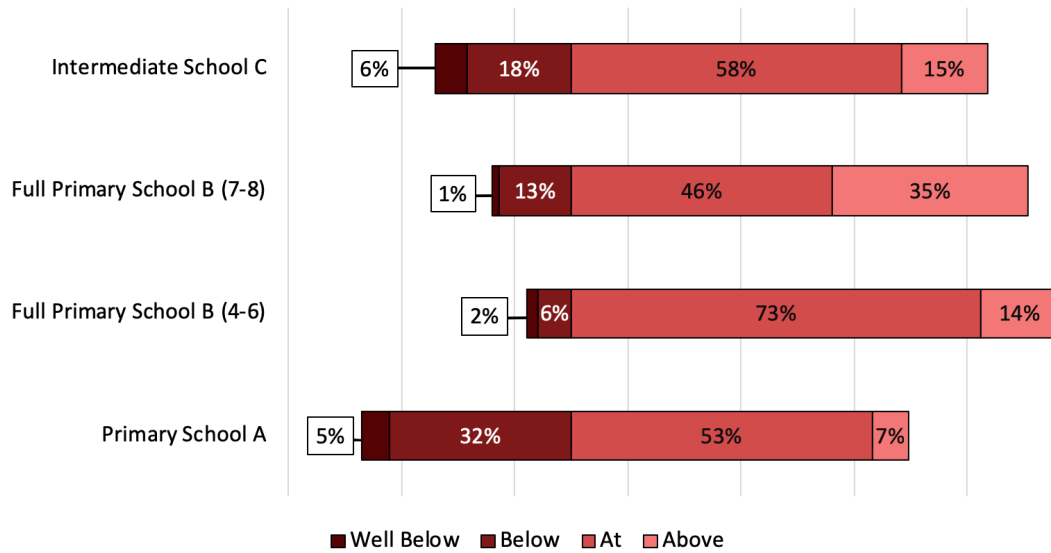


Figure 9.3: Students Reading and Maths levels at each school.

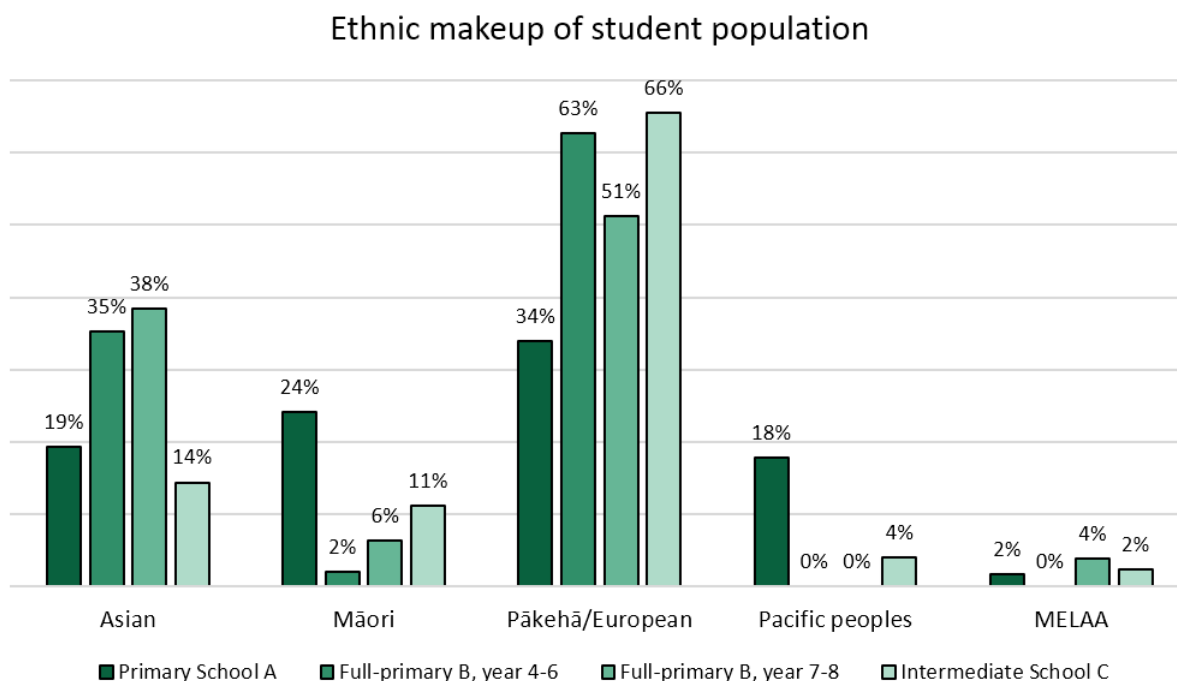


Figure 9.4: Proportion of students belonging to each ethnicity, at each school.

than at Full-primary School B, for both the Full-primary School B year 4-6 group (8%) and the year 7-8 group (16%). This was also true for Maths, where the proportion of students in the Well Below and Below groups for Maths was much higher at Intermediate School C (24%) and at Primary School A (particularly high at 37%) than at Full-primary School B, for both the Full-primary School B year 4-6 group (8%) and the year 7-8 group (14%).

There were also large differences between the ethnic make-up of students at each school, as shown in Figure 9.4. Primary School A had a much higher proportion of Māori and Pacifica students and lower proportion of Pākehā/European students, compared to the other schools. Full-primary School B had a much higher proportion of Asian students, compared to Primary School A and Intermediate School C.

9.3.2 Content covered

Table 9.2 shows what topics each teacher taught. For reference, the required CS Unplugged activities were: Binary numbers, binary representation of text, the Parity card and Barcode tricks for teaching error detection and correction algorithms, and unplugged programming

with KidBots. The Sorting Network and the image representation activities (from the Classic CS Unplugged website) were optional extra lessons. All classes did the required CS Unplugged lessons, except for one who did not do the barcodes activity. Four teachers taught either image representation or the Sorting Network as an additional activity.

9.4 Results: Computational Thinking Tests

The distributions of students' scores in the Castors and Benjamins tests are shown in figures 9.5a and 9.5b. The distribution of scores within each school are shown in figures 9.6a through 9.6d.

I used an independent samples t-test to establish whether the 2016/2017 tests, for each age group, were of similar difficulty, by comparing the pre-test scores of students. The results of this for the Castors (years 4-6) tests are shown in table 9.3, and the results for the Benjamins (year 7-8) tests are shown in table 9.4. These showed that the Castors A and Castors B tests were comparable, but the Benjamins A and Benjamins B tests were not. Because of this, the order year 7-8 students took the tests in will have impacted the change in their scores, so these two groups will be examined separately.

I used a t-test to establish whether there were significant differences between students' pre-test scores depending on the school they attended. Because of the differences between the Benjamins A and B tests, the year 7-8 students are examined separately based the order of tests they took. These showed there was a significant differences between students' pre-test scores at Primary School A and at Full-primary School B (year 4-6), shown in table 9.5. There was also a significant difference between students' pre-test scores, for both Benjamins tests, at Full-primary School B (year 7-8) and at Intermediate School C, as shown in tables 9.6 and 9.7.

Due to these significant differences between the difficulty of some tests, and the significant difference between students' scores at different schools, the following analysis is divided into six sections. These are:

1. Results for Primary School A (years 4-6).
2. Results for Full-primary School B (years 4-6).
3. Results for Full-primary School B (years 7-8), test order A-B.
4. Results for Full-primary School B (years 7-8), test order B-A.
5. Results for Intermediate School C (years 7-8), test-order A-B.

Topics covered					
	Required activities	<i>Sorting Network</i>	<i>Image representation</i>	Python (Code Avengers)	Scratch
Primary School A					
Class A1	Yes	No	Yes	No	Yes (students playing)
Class A2	Yes	Yes	No	No	No
Full-primary					
School B					
Class B1 (4-6)	Yes	Yes	No	No	Yes (Code Club and teacher directed)
Class B2 (4-6)	Yes	No	No	No	Yes (Code Club)
Class B3 (7-8)	Yes	No	No	No	Yes (Code Club)
Class B4 (7-8)	Yes	No	Yes	No	Yes (Code Club)
Intermediate					
School C					
Class C1	Yes	No	No	Yes (2-3 lessons)	No
Class C2	Yes	No	No	Yes (2-3 lessons)	No
Class C3	Yes	No	No	Yes (2-3 lessons)	No
Class C4	Yes, except barcodes	No	No	Yes (6-7 lessons)	No
Class C5	Yes	No	No	Yes (with extension students only)	No

Table 9.2: Topics taught, and programming lessons done in each participating class

Test	N	Mean	Std. Dev	t	Sig.
Castors 2016	53	37.8%	23.8%	0.320	0.750
Castors 2017	60	36.4%	19.2%		

Table 9.3: Comparing difficulty levels of the Castors 2016/2017 (A/B) tests, based on students pre-test scores.

Test	N	Mean	Std. Dev	t	Sig.
Benjamins 2016	100	54.5%	33.3%	4.952	<0.001
Benjamins 2017	103	39.6%	30.7%		

Table 9.4: Comparing difficulty levels of the Benjamins 2016/2017 (A/B) tests, based on students pre-test scores.

School	N	Mean	Std. Dev	t	Sig.
Primary School A	62	26.7%	16.6%	-6.527	<0.001
Full-primary School B (year 4-6)	51	49.6%	19.9%		

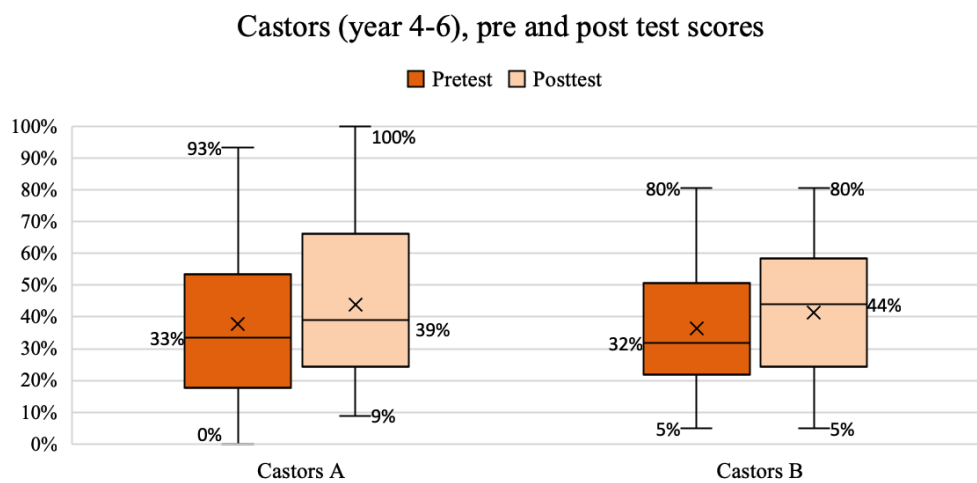
Table 9.5: Comparing results of the Castors tests within different schools, based on the pre-test results.

School (2016 test)	N	Mean	Std. Dev	t	Sig.
Full-primary School B (year 7-8)	37	61.1%	20.7%	-2.269	0.026
Intermediate School C	63	50.7%	24.3%		

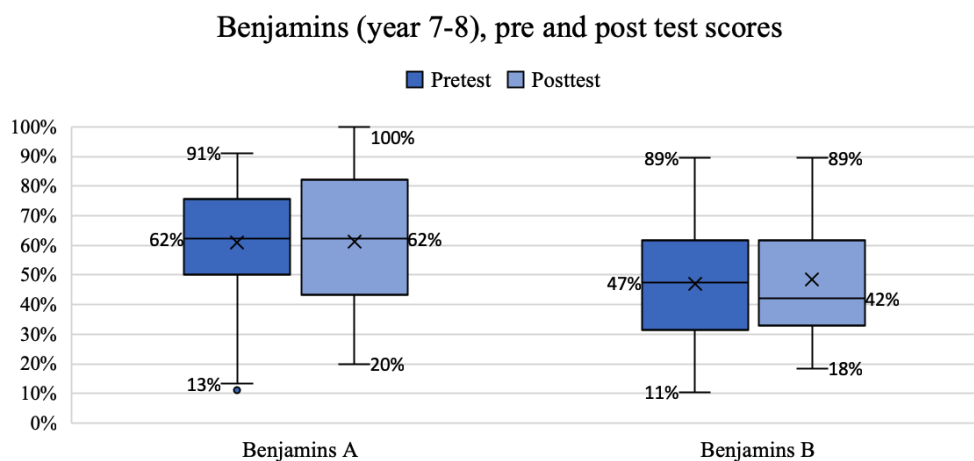
Table 9.6: Comparing results of the Benjamins **A** test within different schools, based on the pre-test results.

School (2017 test)	N	Mean	Std. Dev	t	Sig.
Full-primary School B (year 7-8)	41	47.0%	21.2%	-3.161	0.002
Intermediate School C	62	34.6%	16.5%		

Table 9.7: Comparing results of the Benjamins **B** test within different schools, based on the pre-test results.



(a)



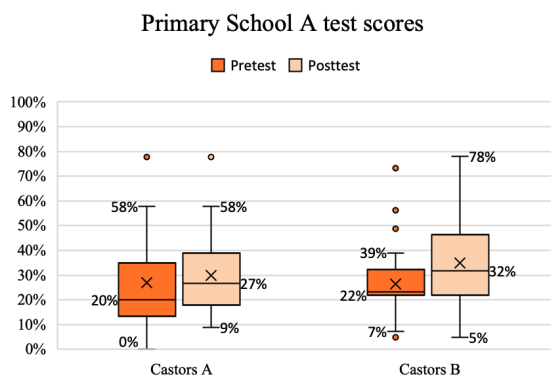
(b)

Figure 9.5: Distribution of scores within each age group

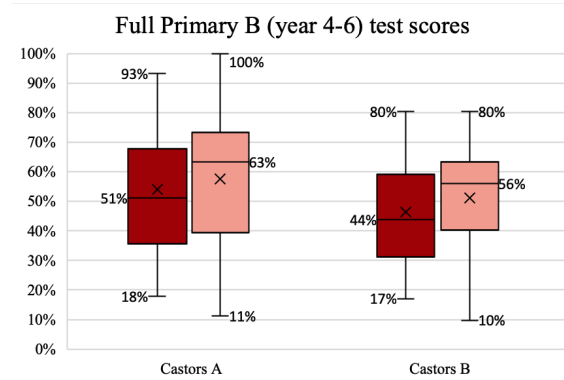
6. Results for Intermediate School C (years 7-8), test-order B-A.

Within each of these sections the following analysis is reported on:

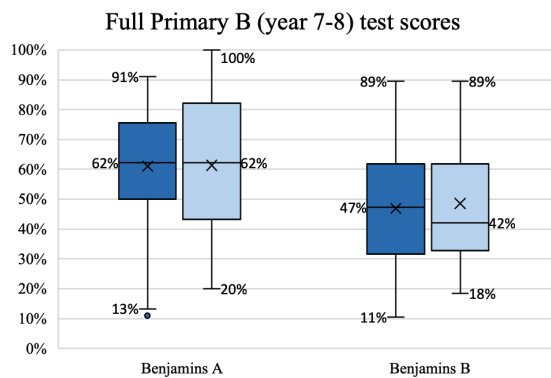
1. Independent samples t-test to compare the learning gain of participants to controls.
2. Independent samples t-test to compare the pre-test scores and learning gain, of female students to male students.
3. Independent samples t-test to compare the pre-test scores and learning gain, of Māori/Pacifica students to non-Māori/Pacifica students.



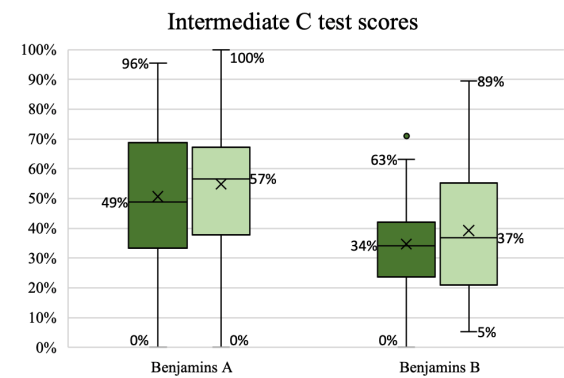
(a)



(b)



(c)



(d)

Figure 9.6: Test scores within each school.

4. One-way ANOVA to compare the pre-test scores and learning gain, of students of different ethnicities.
5. Spearman's correlation to test for a relationship between students' Maths Levels, and their pre-test scores and learning gain.
6. Spearman's correlation to test for a relationship between students' Reading Levels, and their pre-test scores and learning gain.

9.4.1 Primary School A

The results for Primary School A are shown in tables 9.8, 9.9, and 9.10.

- There was no significant difference in learning gain between participants and controls, at the $p > 0.05$ level (Table 9.8).
- There was a significant difference in pre-test scores between female students and male students, at the $p < 0.05$ level (Table 9.9). This was unexpected, due to the body of research showing that gender and sex do not have a significant impact on abilities in maths, programming, and CS. At this school there was a much larger proportion of male students in the 'At' and 'Above' levels in both Maths (85.7%) and Reading (85.7%) compared to the female students (48.7% and 61.5%). Maths and Reading were both found to have a significant impact on students scores in multiple cases, so it is likely that this impacted the difference in female and male students' scores. This difference in Reading and Maths Levels between female and male students was not observed at any other school, and does not reflect the national trends for these levels. Out of all students in NZ, between 75-80% of both female and male students are placed at the 'At' or 'Above' levels for Reading and Maths. Therefore these results on the influence of gender are not representative. There was no significant difference in learning gain between female students and male students, at the $p < 0.05$ level (Table 9.8).
- There was no significant difference in pre-test scores between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.9). There was no significant difference in learning gain between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.8).

Learning Gain, Primary School A					
Group	N	Mean	Std. Dev	t	Sig.
Participant	49	-0.430%	30.39%	0.756	0.453
Control	13	6.82%	32.11%		
Female students	39	-0.70%	27.59%	-0.918	0.362
Male students	21	6.80%	34.60%		
Māori/Pacifica students	26	-2.06%	32.59%	0.685	0.495
Non-Māori/Pacifica students	36	3.37%	29.40%		

Table 9.8: T-test results comparing learning gain within groups of students, at Primary School A.

Pre-test scores, Primary School A					
Group	N	Mean	Std. Dev	t	Sig.
Female students	39	23.07%	12.47%	-2.484	0.016
Male students	21	33.92%	21.42%		
Māori/Pacifica students	26	27.68%	13.65%	-0.379	0.706
Non-Māori/Pacifica students	36	26.04%	18.67%		

Table 9.9: T-test results comparing pre-test results within groups of students, at Primary School A.

Reading and Maths levels, Primary School A				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	60	0.340	0.008
	Learning Gain	60	0.022	0.868
Reading	Pre-test score	60	0.489	<0.001
	Learning Gain	60	0.025	0.848

Table 9.10: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores and learning gain, at Primary School A.

- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(3,55) = 1.25$, $p = 0.30$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(3,55) = 0.46$, $p = 0.71$].
- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Maths and learning gain, at the $p < 0.05$ level (Table 9.10).
- There was a moderately positive correlation between Reading level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Reading and learning gain, at the $p < 0.05$ level (Table 9.10).

9.4.2 Full-primary School B, years 4-6

The results for Full-primary School B, years 4-6, are shown in tables 9.11, 9.12, and 9.13.

- There was no significant difference in learning gain between participants and controls, at the $p < 0.05$ level (Table 9.11).
- There was no significant difference in pre-test scores between female students and male students, at the $p < 0.05$ (Table 9.12). There was a significant difference in learning gain between female students and male students, at the $p < 0.05$ level (Table 9.11).

Learning Gain, Full-primary School B(4-6)					
Group	N	Mean	Std. Dev	t	Sig.
Participant	34	3.06%	38.70%	1.335	0.188
Control	17	18.56%	39.96%		
Female students	23	21.92%	35.27%	2.348	0.023
Male students	28	-3.02%	39.66%		

Table 9.11: T-test results comparing learning gain within groups of students, at Full-primary School B (year 4-6 students).

Pre-test scores, Full-primary School B(4-6)					
Group	N	Mean	Std. Dev	t	Sig.
Female students	23	48.03%	19.53%	-0.500	0.619
Male students	28	50.84%	20.42%		

Table 9.12: T-test results comparing pre-test results within groups of students, at Full-primary School B (year 4-6 students).

Reading and Maths levels, Full-primary School B (year 4-6 students)				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	48	0.340	0.018
	Learning Gain	48	0.284	0.050
Reading	Pre-test score	48	0.168	0.254
	Learning Gain	48	0.247	0.091

Table 9.13: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores and learning gain, at Full-primary School B (years 4-6 students).

This may have been due to their being more female students in the B-A test group than there were in the A-B group, while the number of male students in each of these groups was relatively even.

- There was only one Māori/Pacific student in this group of students, so no comparisons could be made between Māori/Pacific and non-Māori/Pacific students.
- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(2,48) = 0.055$, $p = 0.95$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(2,48) = 1.476$, $p = 0.24$].
- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.05$ level. There was a weak positive correlation between Maths and learning gain, which was marginally statistically significant at the $p < 0.05$ level, with $p = 0.050$ (Table 9.13)
- There was no significant correlation between Reading and pre-test score, at the $p < 0.05$ level. There was no significant correlation between Reading and learning gain (Table 9.13).

9.4.3 Full-primary School B, years 7-8, test order A-B

The results for Full-primary School B, years 7-8 and test order A-B, are shown in tables 9.14, 9.15, and 9.16.

- There was no significant difference in learning gain between participants and controls, at the $p < 0.05$ level (Table 9.14).
- There was no significant difference in pre-test scores between female students and male students, at the $p < 0.05$ (Table 9.15). There was no significant difference in learning gain between female students and male students, at the $p < 0.05$ (Table 9.14).
- There was only one Māori/Pacific student in this group of students, so no comparisons could be made between Māori/Pacific and non-Māori/Pacific students.

Learning Gain, Full-primary School B (7-8, A-B)					
Group	N	Mean	Std. Dev	t	Sig.
Participant	22	-12.14%	24.19%	1.119	0.271
Control	15	-22.77%	33.69%		
Female students	15	-19.35%	28.46%	-0.506	0.616
Male students	22	-14.47%	28.98%		

Table 9.14: T-test results comparing learning gain within groups of students (between tests A-B), at Full-primary School B (years 7-8 students).

Pre-test scores, Full-primary School B (7-8, A-B)					
Group	N	Mean	Std. Dev	t	Sig.
Female students	15	63.56%	23.07%	0.594	0.556
Male students	22	59.39%	19.33%		

Table 9.15: T-test results comparing pre-test results within groups of students (test A), at Full-primary School B (years 7-8 students).

Reading and Maths levels, Full-primary School B (7-8, A-B)				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	36	0.492	0.002
	Learning Gain	36	-0.227	0.184
Reading	Pre-test score	37	0.603	<0.001
	Learning Gain	37	-2.00	0.236

Table 9.16: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores (test A) and learning gain (between tests A-B), at Full-primary School B (years 7-8 students).

- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(3,33) = 1.010$, $p = 0.40$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(3,33) = 0.512$, $p = 0.68$].
- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Maths and learning gain, at the $p < 0.05$ level (Table 9.16).
- There was a moderately positive correlation between Reading level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Reading and learning gain, at the $p < 0.05$ level (Table 9.16).

9.4.4 Full-primary School B, years 7-8, test order B-A

The results for Full-primary School B, years 7-8 and test order B-A, are shown in tables 9.17, 9.18, and 9.19.

- There was no significant difference in learning gain between participants and controls, at the $p < 0.05$ level (Table 9.17).
- There was no significant difference in pre-test scores between female students and male students, at the $p < 0.05$ level (Table 9.18). There was no significant difference in learning gain between female students and male students, at the $p < 0.05$ level (Table 9.17).
- There was no significant difference in pre-test scores between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.18). There was no significant difference in learning gain between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.17).
- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(3,37) = 0.635$, $p = 0.60$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(3,37) = 1.230$, $p = 0.31$].

Learning Gain, Full-primary School B (7-8, B-A)					
Group	N	Mean	Std. Dev	t	Sig.
Participant	19	30.49%	42.34%	0.193	0.848
Control	22	28.31%	29.77%		
Female students	19	33.80%	31.88%	0.743	0.462
Male students	22	25.45%	38.99%		
Māori/Pacifica students	4	1.67%	11.64%	-1.668	0.103
Non-Māori/Pacifica students	37	32.31%	36.18%		

Table 9.17: T-test results comparing learning gain within groups of students (between tests B-A), at Full-primary School B (years 7-8 students).

Pre-test scores, Full-primary School B (7-8, B-A)					
Group	N	Mean	Std. Dev	t	Sig.
Female students	19	46.95%	18.30%	-0.009	0.993
Male students	22	47.01%	23.69 %		
Māori/Pacifica students	4	42.76%	28.70%	-0.417	0.679
Non-Māori/Pacifica students	37	47.44%	20.59%		

Table 9.18: T-test results comparing pre-test results within groups of students (test B), at Full-primary School B (years 7-8 students).

Reading and Maths levels, Full-primary School B (7-8, B-A)				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	38	0.534	0.001
	Learning Gain	38	0.278	0.092
Reading	Pre-test score	41	0.424	0.006
	Learning Gain	41	0.441	0.004

Table 9.19: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores (test B) and learning gain (between tests B-A), at Full-primary School B (years 7-8 students).

- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Maths and learning gain, at the $p < 0.05$ level (Table 9.19).
- There was a moderately positive correlation between Reading level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was a moderately positive correlation between Reading level and learning gain, which was statistically significant at the $p < 0.01$ level (Table 9.19).

9.4.5 Intermediate School C, test-order A-B

The results for Intermediate School C, test order A-B, are shown in tables 9.20, 9.21, and 9.22.

- There was no significant difference in learning gain between participants and controls, at the $p < 0.05$ level (Table 9.20).
- There was no significant difference in pre-test scores between female students (and male students, at the $p < 0.05$ level (Table 9.21). There was no significant difference in learning gain between female students and male students, at the $p < 0.05$ level (Table 9.20).
- There was no significant difference in pre-test scores between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level Table 9.21. There was

Learning Gain, Intermediate School C (A-B)					
Group	N	Mean	Std. Dev	t	Sig.
Participant	47	-24.57%	33.32%	-0.801	0.426
Control	16	-16.99%	30.71%		
Female students	29	-16.95%	32.77%	1.577	0.120
Male students	33	-29.57%	30.26%		
Māori/Pacifica students	12	-33.39%	30.86%	-1.275	0.207
Non-Māori/Pacifica students	51	-20.12%	32.78%		

Table 9.20: T-test results comparing learning gain within groups of students (between tests A-B), at Intermediate School C.

Pre-test scores, Intermediate School C (A-B)					
Group	N	Mean	Std. Dev	t	Sig.
Female students	29	48.97%	21.98%	-0.368	0.714
Male students	33	51.25%	26.22%		
Māori/Pacifica students	12	44.63%	17.92%	-0.959	0.341
Non-Māori/Pacifica students	51	52.11%	25.51%		

Table 9.21: T-test results comparing groups pre-test scores at Intermediate School C (test A).

Reading and Maths levels, Intermediate School C (A-B)				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	62	0.586	<0.001
	Learning Gain	62	0.083	0.520
Reading	Pre-test score	63	0.532	<0.001
	Learning Gain	63	0.207	0.107

Table 9.22: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores (test A) and learning gain (between tests A-B), at Intermediate School C.

no significant difference in learning gain between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.20).

- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(4,57) = 0.223$, $p = 0.92$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(4,57) = 0.561$, $p = 0.69$].
- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Maths and learning gain, at the $p < 0.05$ level (Table 9.22).
- There was a moderately positive correlation between Reading level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Reading and learning gain, at the $p < 0.05$ level (Table 9.22).

9.4.6 Intermediate School C, test-order B-A

The results for Intermediate School C, test order B-A, are shown in tables 9.23, 9.24, and 9.25.

- There was no significant difference between learning gain between participants and controls, at the $p < 0.05$ level (Table 9.23).

Learning Gain, Intermediate School C (B-A)					
Group	N	Mean	Std. Dev	t	Sig.
Participant	47	28.38%	38.46%	-0.098	0.923
Control	15	29.46%	33.39%		
Female students	37	27.14%	34.08%	-0.468	0.642
Male students	23	31.78%	42.01%		
Māori/Pacifica students	7	43.24%	20.77%	1.110	0.272
Non-Māori/Pacifica students	55	26.78%	38.34%		

Table 9.23: T-test results comparing learning gain within groups of students, at Intermediate School C (between tests B-A).

Pre-test scores, Intermediate School C (B-A)					
Group	N	Mean	Std. Dev	t	Sig.
Female students	37	35.49%	17.01%	0.413	0.680
Male students	23	33.64%	16.70%		
Māori/Pacifica students	7	23.68%	8.46%	-1.903	0.062
Non-Māori/Pacifica students	55	36.03%	16.81%		

Table 9.24: T-test results comparing groups pre-test scores at Intermediate School C (test B).

Reading and Maths levels, Intermediate School C (B-A)				
		N	Corr. Coefficient	Sig.
Maths	Pre-test score	60	0.402	0.001
	Learning Gain	60	0.227	0.80
Reading	Pre-test score	62	0.401	0.001
	Learning Gain	62	0.181	0.167

Table 9.25: Spearman's correlation between students' Maths and Reading levels, and their pre-test scores (test B) and learning gain (between tests B-A), at Intermediate School C.

- There was no significant difference in pre-test scores between female students and male students, at the $p < 0.05$ level (Table 9.24). There was no significant difference in learning gain between female students and male students, at the $p < 0.05$ level (Table 9.23).
- There was no significant difference in pre-test scores between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.24). There was no significant difference in learning gain between Māori/Pacifica students and non-Māori/Pacifica students, at the $p < 0.05$ level (Table 9.23).
- A one-way between subjects ANOVA was conducted to compare the effect of students' ethnicity on pre-test scores, and another to compare the effect of ethnicity on learning gain. There was no significant impact of ethnicity on students' pretest scores at the $p < 0.05$ level [$F(4,55) = 1.092$, $p = 0.37$]. There was also no significant impact of ethnicity on learning gain at the $p < 0.05$ level [$F(4,55) = 0.712$, $p = 0.59$].
- There was a moderately positive correlation between Maths level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Maths and learning gain, at the $p < 0.05$ level (Table 9.25).
- There was a moderately positive correlation between Reading level and pre-test score, which was statistically significant at the $p < 0.01$ level. There was no significant correlation between Reading and learning gain, at the $p < 0.05$ level (Table 9.25).

9.4.7 *Summary*

No cases were found where students who participated in these classes improved more than the control group. Therefore, the null hypothesis “participating in these lessons has no significant impact on students’ CT skills” cannot be rejected. Students’ gender and ethnicity were not shown to have an impact on students’ results, except for one case which showed a correlation between gender and achievement. This correlation was likely to be invalid however, due to the variation in Reading and Maths levels between female and male students in this sample. Read and Maths levels were frequently found to impact students’ results. These results and the limitations on their validity are discussed in section 9.6.

9.5 **Results: Interviews**

9.5.1 *Teachers*

I worked with three teachers at Primary School A, four at Full-primary School B, and six at Intermediate School C. Unlike my previous two studies, these teachers were not entirely self-selected. At each school, one of the teachers had volunteered their school for the study, but others at the school had been asked to join so that their school could participate.

In the interviews, I asked teachers how they had felt at the beginning and the end of the study. There was a clear divide between teachers who felt anxious and stressed at the beginning and those who did not. The teachers who had some prior experience or exposure to CS or programming felt generally excited and motivated. Those who did not, described themselves as “terrified”, “nervous”, and said they “didn’t feel confident at all”. Some of these teachers commented that after the workshop on teaching CS Unplugged they felt more excited and reassured. Several of these teachers, and also one of the more experienced and confident teachers, felt stressed or un-enthused about trying to fit the lessons into their timetable. However, each of these teachers said that in the end, it was worth finding time for these:

- *Before they began teaching this* — “I was quite excited. I was buzzing, but I was anxious about everything. I was anxious about all of the lessons and getting everything done in the time frame that I had allocated.”

At the end of the study, when asked if it was worth teaching — “Absolutely. I wouldn’t be rolling out over the whole school if I didn’t think it was.”

- “I was quite nervous. In my head it was just another thing that I had to try and squeeze into my day. But we did it and it was worth doing. After the first couple of lessons I came out feeling okay about it.”
- *Before they began teaching this* — “I was a bit hesitant because I was thinking ‘Oh no, this is another thing we have to try and do, when are we going to fit this in?’ ... there was a little bit of worry... It probably wasn’t really an area of interest for me. I know it’s kind of the age we are coming into and a lot of our students may end up with jobs in that area, but it still seemed to me like this was still a bit further away, not ages away but still a bit in the distance.”

At the end of the study — “I now feel like I’m a step ahead of a lot of people... That has been a positive. I’m still trying to get my head around it a bit more. Of course, now I know a lot more about it now than I did before, but it still feels a bit complex, which I guess it always is because it’s new stuff and it’s not easy... I would really like to teach it again.”

At the end of the study, all teachers were glad that they had taken part, regardless of their initial confidence and motivation levels. Teachers who were initially more anxious and lacked confidence had become more comfortable with the teaching materials, but some felt it would take more time before they were confident with the subjects. Several said one of the positive outcomes for them was they now felt they had a “head start” for when the new curriculum came in. Some examples of comments on these positive outcomes for the teachers themselves are:

- “I think knowing that it’s going to become something that we have to teach, I feel better that I’ve jumped the gun and done it earlier.”
- “it’s good to get a head start, and I like [the subject].”
- “we were all a bit nervous because.. But I’m feeling okay about it this year, now I’ve done it once.”

Many teachers discussed their experiences of learning alongside their students, as they were just as new to the subjects as them. As in the previous study, they saw this as a positive part of students’ learning, and that they had fun learning with their students:

- “It was a really nice opportunity to tell the children that I was actually learning as well. I think, hopefully, they would have respected that. There were times when things

didn't go quite right so would say to them 'Wait, hang on a minute, I just need to go look again!' So, it was actually teaching the children that everyone learns."

- "I guess the beauty of when you are only a couple of steps ahead of the kids is that you still understand how challenging a concept it is to understand."
- "I loved learning alongside with them as well. I loved doing it with them"
- "I was literally reading from the script. We were learning together, and they thought that was kind of cool as well because I didn't know what I was talking about, so we sort of got it together... I think then seeing me not knowing what I was talking about was also beneficial for them, just to know that I am only human as well, and we don't all love every subject."

While these experiences are related to this being a new subject area, rather than being specific to CS and programming, they can offer insights into how other teachers could be enthused to take up this subject in the future.

9.5.2 Lessons taught

For this study I asked teachers to teach the CS Unplugged activities and programming as stand-alone lessons, and if they did any integration with other subjects to do this after they had first taught them in isolation. I did this as I was investigating the specific link between only CS and programming, and the development of CT skills. In my previous studies, the majority of teachers had done a degree of integration between these topics and other subjects, which had been highly successful. I was now aiming to find if this success could be (at least partially) attributed to the CS and programming content, or if it was only due to the manner in which these were taught alongside other subjects. One teacher integrated Scratch programming with a unit they did on Growth Mindset, and after the end of the study another began integrating Scratch into other subjects. Other than this, all teachers who taught programming taught it as a stand-alone topic. One teacher did a small amount of integration between the barcode activity and maths, but only in the sense that they used the activity to motivate students who were not hugely engaged with maths, to work on their numeracy skills. One teacher taught this as a part of their maths time but made it clear to students that 'coding' (how they referred to these activities and programming) was a separate topic to numeracy (their usual maths programme). Apart from these small links

with the maths curriculum, the CS Unplugged activities were taught as stand-alone topics by all teachers.

All teachers stated that these lessons had been beneficial for their students, the CS Unplugged activities had been particularly successful, and the resources for these had made teaching it simple. When asked if there were any negative impacts on their students, teachers' answers were akin to those given in the prior study. The majority of teachers said they observed no negative impacts on the students involved, and one said the only students there was a negative for were those who could not take part (due to not receiving parental permission to participate). Similarly, one teacher said the only negative was that their students wanted to share what they had learnt with their friends in other classes, but were not able to as these students were a part of the control group:

“I know that the kids were really keen to share their learning, they actually started writing in posts about it, which I encouraged them to do but I had to ask them not to post them yet and they had to wait until the end of term. Which was really hard for them and a shame, but you could see that as a positive as well because they really wanted to share their learning.”

Two teachers cited frustration as the only potential negative impact on their students. For one class this was frustration during the Bebras pre-test, and their teacher felt this initially lowered students confidence. However, they went on to say “they got over that pretty quickly, so I don’t think it negatively impacted them in any way”. The remaining teacher’s students, on the other hand, did not recover from this frustration: “For certain students, they got so angry! Their frustration was just at explosion level... They are the ones who gave up... The frustration was too much for them”. The teacher also said, “ But that’s just their personalities”.

Computer Science Unplugged

Teachers said the activities were engaging and enjoyable for all, or the vast majority, of their students. Several teachers said there were a small number of students (generally two to five) who were not significantly engaged with the material. The reasons given for this included some students approaching the activities with “an element of arrogance” and viewing the activities as too “babyish” for them, or students having very high learning needs. The majority of teachers also described how there were groups of students in their classes who

were exceptionally interested in the topics, and “ it certainly pushed some kids’ buttons more than others”.

When asked, every teacher stated that the CS Unplugged activities were beneficial for their students and highly successful. Some examples of how teachers described the general success of the lessons are:

- “They loved it. They got really good results... [the activities] were awesome. The directions were really clear, and the teacher support stuff was awesome. The worksheets that went along with it were great, because they enjoyed them and I enjoyed it.”
- “[The activities] were successful because they made them think. The kids loved it and they all got into it. That’s why they were good. And it got them thinking as well as them having fun. They talked a lot, especially during the KidBots activity. They planned together how to make the activity more challenging. That was really good”
- “Everything in the unplugged went really well... I think on the whole the unplugged lessons were super positive and I think that they took away a really different view of what computer science is, and how computers work”
- When asked if the lessons were successful —“Definitely”, “Yes, the children definitely enjoyed it, which was really good.”
- “I think for most kids, they were all really engaged regardless of their level. Whether they were an above, at, or a below student, they could all do it. Especially because most of the stuff was done in little groups, and they would go with someone who could help them and they all helped each other.”

When asked if all their students had been able to attain some level of success in these lessons the majority of teachers said yes, and the others said the majority of their students had. In the cases where not all students were able to achieve to some degree, teachers said it was because these students were not engaged. These were the same group of disengaged students as described previously. Another teacher had a group of students who became extremely frustrated with the material and gave up, though it was not clear if this was in relation to the CS Unplugged material, the Python programming, or both. They also commented that this behaviour was not unexpected for these students. One teacher also found that during the KidBots activity one group of students struggled more than others:

“I know for KidBots a few struggled. One group, who were more of the low achievers who struggle to work with other people sometimes, they did struggle with the group

stuff and trying to give instructions to each other... and not realising that if you didn't get it right then you had to go all the way back to the start... The kids who don't have the best social skills did struggle with a few of the lessons. But that wasn't a big deal because you either just put them in a different group or I hung around them and helped them with what we were doing."

Along with the teachers' opinions on the general success of the lessons, I was interested in whether they felt students had developed knowledge of actual CS concepts. Much of the feedback received on the CS Unplugged lessons, throughout all three studies, was focused on the high engagement of students, and that the tactile 'hands-on' nature of the activities was one of the students' favourite things about these lessons. I asked teachers if they thought their students had gained an understanding of the concepts the lessons were intended to teach, rather than *just* enjoying them. Three teachers were not sure if students had seen the concepts behind some of the activities, but it was not entirely clear from their interviews whether or not their students had gained at least some CS knowledge. One of these teachers commented that when students did the CS Unplugged activities they thought their students probably "didn't see it as learning. I don't think they realised and they just thought it was fun". One teacher stated that their students had not gained an understanding of the underlying concepts in the parity trick activity, and they had seen this as "just a magic trick". Apart from these cases however, teachers said their students almost certainly had learnt about the CS concepts the activities were intended to teach and had gained an understanding of how these related to their everyday lives:

- "Yes, definitely... I was really surprised at how quickly they grasped the concepts."
- "Absolutely... explaining how the concepts connect to a bigger picture helps them, especially at the year eight level, where they can kind of conceptualise what the next steps are. I think it worked well."
- 'Yes, definitely... I think they could see the connections between maths and what they were doing, but also how things work. Like 'Oh, so that's how checkouts work, and things come off the inventories of the shop' so they could actually really make a connection with how it works"
- *Teacher*: "It was quite neat to see when they actually started to get it, when they had those lightbulb moments when they started understanding it, because at first it is a little bit confusing because it is all new... I obviously talked about how computers are run by codes and that the codes, in its simplest sense is really just off and on. I think

they were sort of starting to get that, and then we talked about the binary numbers and how we can change them, and I think they were starting to cotton onto that.”

Interviewer: “Cottoning on to those deeper computer science concepts?”

Teacher: “Yes.”

- “with the binary stuff, yes definitely... They definitely got it well the kid bots as well. Those were the two main ones where they really had to change their thinking.”

Programming

I had originally intended for all teachers to use the Code Avengers platform for programming lessons. At the time when the study was planned the courses available through Code Avengers were all targeted at High School students, but I chose to use this platform as Junior courses were planned to be released before the study began. However, these lessons were subsequently not released in time to be used in this study. Because of this, teachers at Primary School A and Full-primary School B did not use Code Avengers. Several classes at Full-primary School B attempted the first activity in the Python course but stopped when they found the reading level was, understandably, too high for the majority of their students. They then decided the Code Club Scratch resources would be more suitable, and each used these for several lessons. Several of the year 7 and 8 students continued with the Python course in their own time. The two teachers interviewed at Primary School A did not do any programming lessons, but in one class a student was particularly interested in Scratch programming so several students experimented with this. It was not clear from their interview how much Scratch programming these students did in class.

All but one teacher at Intermediate School C had their students spend two or more lessons on the Code Avengers Python programming course. The remaining teacher at Intermediate School C had four of their extension students try these lessons out, and these students continued with the lessons in their own time. The Python resources were not particularly successful, which was unsurprising considering the target age group of these. Three of the four teachers who had their whole class try the Python course commented that the majority of their students were able to complete the first two or three activities, but stopped after that as the reading became too challenging and they lost interest. One of these teachers commented that, although the majority of students did not complete much of the Python course, they felt there had been some benefit to most students as they had gained some familiarity with debugging.

On the other hand, one class at Intermediate School C did continue with the Python course, and spent around six or seven lessons on these. The teacher for this class had experience with Computer Science and Scratch programming before the study began, and so may have been able to assist their students more with these programming activities. This teacher still felt these lessons were not ideal for this age group and Scratch would have been a better option, but their students did still have success. When they were asked if they thought their students had gained an understanding of programming from these lessons they said:

“They got the idea... that coding actually is getting a sequence of instructions, but at times you have to debug it to make it work. They got that concept quite clearly and were able to do the debugging”.

The teachers at Full-primary School B who used the Scratch resources from Code Club found these were generally successful and beneficial for students. When asked if the lessons were successful one teacher answered “Yes. The whole lot was, the Unplugged and the programming”, and another described the programming lessons as successful and added “I actually think it was more powerful as we had done those previous [CS Unplugged] lessons”. Another teacher at Full-primary School B had since begun integrating programming into many parts of their classroom programme, as a way to engage students with their learning. They found this was beneficial in other curriculum areas and, like in the previous study, said it enabled students to present their learning in a new way:

“I try to involve coding into lots of different things as a gaming type, or enjoyment part of learning. I think that if we get them to do something... like in ‘chapter chat’ over there [*gestures to students’ reading and writing work on the wall in the classroom*]. They have to visualise what’s happening in the story so I would say ‘why don’t you make the sequence of what is happening in the story on Scratch?’ Because they can visualise it in their head, but that articulation of it is really difficult... But if they can show me, I can understand.”

9.5.3 Computational Thinking

As in the previous Open Teachers Study, I reviewed the interview transcripts to identify cases where students were exercising CT skills. The same criteria (listed in Chapter 8, section 8.2.2) as in the previous study were used for tagging these observations. - tables all changed

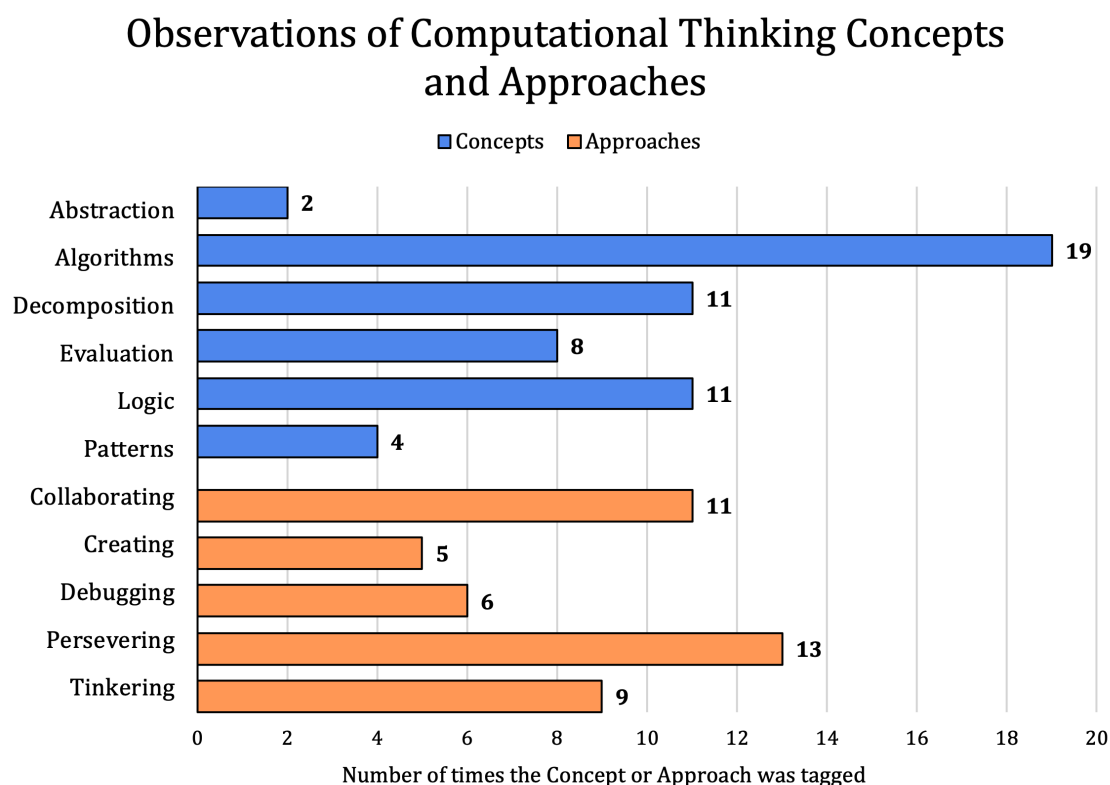


Figure 9.7: The Computational Thinking Concepts and Approaches, which were identified in the interview transcripts.

The proportion of teachers who observed each skill in this study was lower for every CT skill when compared with the previous study. These differences are shown in table 9.27. While the differences were relatively small for Abstraction, Algorithms, Collaboration, and Persevering, they were large (between -30% and -64%) for every other skill. This indicates students' CT skills may have developed significantly less in this study. There are several possible explanations for this, which are discussed in section 9.6.

Algorithms and Algorithmic Thinking, Decomposition, Evaluation, Logic, and Debugging

These four skills have been grouped as each was almost always tagged alongside one or more of the others. Students were observed learning about, or demonstrating selections of these skills through learning and carrying out algorithms; comparing different algorithms;

Out of a total of 11 teachers			
CT Skill	Total observations	Number who observed the skill	Proportion who observed the skill
Concepts			
Abstraction	2	1	9%
Algorithms	19	10	91%
Decomposition	11	7	64%
Evaluation	8	5	45%
Logic	11	7	64%
Patterns	4	2	18%
Approaches			
Collaboration	11	8	73%
Creating	5	4	36%
Debugging	6	4	36%
Persevering	13	8	73%
Tinkering	9	5	45%

Table 9.26: The total number of observations of each Computational Thinking Concept and Approach, and the number of teachers that observed each of these.

identifying and fixing mistakes made when carrying out algorithms; through the KidBots activity as they worked on breaking down a process into steps, identifying and fixing bugs, extending the activity by creating new challenges, and comparing the efficiency of their solutions; and through writing programs, identifying problems in these, and debugging them. Some examples of sections which were tagged with two or more of these skills are:

- “I do feel that there was more of an understanding of how to get from one place to another, and that you had to do things in a sequence, and if you didn’t do them in sequence then perhaps things wouldn’t work the way you expected them to. It was all about planning, logical thinking, logical reasoning, and justifying your choices.” — **Algorithms, Decomposition, Evaluation, Logic, and Debugging**
- “They got the idea, especially with the one with the magnets, and that coding actually is making a sequence of instructions, but at times you have to debug it to make it work. They got that concept quite clearly and were able to do the debugging” — **Algorithms, Decomposition, and Debugging**
- “I could see them thinking that they really needed to give the step-by-step specific directions. They were like ‘come on you just go from there to there to there’, and I said,

Proportion of teachers observing each CT Skill			
CT Skill	Open teachers study (interviewed only)	Intervention study	Difference
Concepts			
Abstraction	17%	9%	-8%
Algorithms	100%	91%	-9%
Decomposition	94%	64%	-30%
Evaluation	89%	45%	-44%
Logic	94%	64%	-30%
Patterns	78%	18%	-60%
Approaches			
Collaboration	78%	73%	-5%
Creating	89%	36%	-53%
Debugging	100%	36%	-64%
Persevering	83%	73%	-10%
Tinkering	89%	45%	-44%

Table 9.27: Differences in proportion of teachers each skill was observed by in interviews, between the Open teachers study and the Intervention study

‘no that actually doesn’t work, you need to be specific’. That specific command thing, they actually enjoyed that.” — **Algorithms, Decomposition, and Debugging**

- Discussing students programming — “they realised ‘oh right I made a mistake, next time I will need to fix that’.” — **Logic, Evaluation, Debugging**
- “They were able to talk about ‘look I do actually have a problem there, what is the problem?’ Kind of talk about how to get around the problem. That was probably more the higher achieving students because they were quite critical in their thinking. They tried to find ways that things didn’t work.” — **Evaluation, Logic, Debugging**

The following discussion with a teacher on how debugging became a normal part of their classroom programme also illustrated all five of these skills:

Teacher — “When one thing doesn’t happen what do you need to do? You need to go back and check it. We say that in everything now, we constantly say in writing ‘go back and check your work, check your mistakes, find the bug’. I use that now in writing. I’ll say, ‘your writing is bugged, let’s go back and find what the problem is’, things like there isn’t a comma or a full stop... When I think about what they were do-

ing with KidBots it's trial and error, and then you find the mistakes and you fix them."

Interviewer — "And they were doing that?"

Teacher — "They were doing it, and they did it all through the rest of the year... they were like 'let's do it again! Let's try and fix it. Let's try and do it as short as we can' or do the shortest possible sequence of steps."

— **Algorithms, Decomposition, Evaluation, Logic, and Debugging**

Along with these observations, Algorithms and Logic were tagged separately several times. The sections which were tagged with Algorithms, and none of these other four skills, either contained explicit references to algorithmic thinking, or were observations of students learning, carrying out, and practising specific algorithms. The sections which were tagged with Logic, and none of these other four skills, were all explicit references to students developing or using logical reasoning skills, with no reference to any of the other skills.

Abstraction

As in the previous study, there were very few observations of Abstraction. The observations of Abstraction were both made by the same teacher and were made during data representation lessons. They described how students learnt about the Parity and Barcode tricks and were then able to see that the overall idea behind these was that extra information can be added to data and then used for error checking. They observed this pattern in both the activities and could see that despite the differences in how the error checking was performed, that both of these activities were about the same concept. The teacher also described how both they and their students were then able to connect their knowledge of data representation, to how all digital technology operates on the concepts of using binary data to represent other types of data, such as videos.

"I think it was quite cool for them to have an insight into the technology they use without even thinking about it every day. And how clever it is, what it breaks down too. Especially when we did binary it's about mind blowing to think, wow I just expect Netflix to work, and when that doesn't or starts slowing down I'm annoyed with it. But when it comes down to it there is a lot going on in that"

These observations were also examples of understanding Decomposition, and Patterns and Generalisation.

Patterns and Generalisation

The observations of Patterns and Generalisation were all related to students recognising and understanding patterns, but no explicit observations of generalisation were found. The recognition and use of patterns occurred in the binary, and error detection and correction activities. One observation was related to the Bebras test, and students general learning about patterns and logic. This may have been referencing students applying their pattern recognition skills across a range of different problems (i.e. generalisation), but it was not entirely clear if this is what the teacher meant, and so this was not tagged.

Some examples of the Patterns and Generalisation observations are:

- “they picked up that there was a purpose to that extra information, and they saw the patterns.”
- “The whole binary thing they just got really fast, like ‘oh yeah, it’s this or this, it’s this or this’.”

Collaborating

Similarly to the previous study, there were many observations of students working together and improving their teamwork skills; students who would not normally work well with other students being more motivated to do this, and working on their social skills by doing this; and students supporting each other in their learning

Some examples of the sections tagged as Collaboration are:

- Discussing students who were generally high achieving —“because they were able to help other students understand concepts, they were building relationships with students in the class that they hadn’t necessarily had a constructive relationship with.”
- “It was good to see students collaboratively working, it was great.”
- “I had kids working with each other who would have never worked with each other before, they were sharing knowledge. That was beneficial.”
- “The kids bounced off each other, so the whole time they would say things like ‘look at our KidBots grid and program!’, ‘Oh you did that, now I want to go and try that’. There was that growth that you always hope for, with this it was obvious that they were building off of the things each other were doing.”

Creating

As previously noted, Creating was observed much less frequently (compared to other CT skills) than it was in the previous study. The drop in the amount of Creating was to be expected, as the CS Unplugged and programming lessons taught did not require students to create original digital or physical artefacts, and the lessons were taught in isolation and according exactly to the lesson plans. Several teachers referenced students using Scratch in their own time, but did not describe programs students may have created, and so these were not tagged as Creating.

The other observations of Creating were of students modifying and extending the KidBots activities, by adding new challenges and features to the activity:

“They figured out how to do more and try and make it more exciting. They put these things in, I think they called them bombs, that you had to try and avoid, and they would try to trap you in, so you would have to go all the way around them.”

Persevering

Multiple teachers referenced students increased persistence in the Bebras post-test compared to the pre-test, and persistence when working with challenging reading requirements when programming. These were not factors in the previous study. Some examples of the sections tagged as Persevering are:

- “They grew in that whole problem solving aspect, and the fail fast concept, and knowing that things might not go right the first time but if they keep persevering they will get there... One of our key characteristics is getting our children to be resilient so it’s a perfect way to teach them about that and about persevering. It links really well with our core values.”
- “When they did the Bebras test again you could actually see a bit more of their resilience and that they were willing to try a bit more.”

When asked to described their class in general one teacher described how many students in their class usually did not persist when things became challenging, but in these lessons the students behaved differently:

Teacher — “Lots of students had no patience and they would just call out that what they were doing was too hard and then the whole class would respond to that

and decide it was too hard. Lots of students actually gave up on things. They didn't really persist."

Interviewer — "Were they like that when they did the CS Unplugged activities?"

Teacher — "No actually. Those were different. That was much more interesting for them."

Tinkering

Tinkering was frequently tagged alongside Evaluation, and these examples were similar to those in the previous study. The main observations of Tinkering were of students experimenting with Scratch; thinking of new ways to extend the KidBots activity; and repeating algorithms from the lessons (such as the Barcode trick) many times in order to show other people, and to look for situations where the algorithm did not work. Some examples of the sections tagged as Tinkering are:

- When describing the Barcode trick — "They were really mind blown by it, because they were saying 'What? How did you get that? That doesn't work'. Especially one boy who didn't normally do a lot of work in class, he was going around asking everyone to do it, and was trying to work out or find a way when it doesn't work as well. He wanted to figure out if it maybe didn't work every time."
- "When we were doing coding, after that they were much more likely to say 'I'll give this a go'."
- "We spent a lot of time on KidBots. They spent a lot of time making more courses to make KidBots harder, things like needing to do concurrent movements."

There were less examples of students employing trial and error strategies in this study, which may be due to the lack of creating programs and then testing these.

9.5.4 Problem Solving

Teachers were each asked if they had observed any improvements or changes in students' problem solving skills, as a result of these lessons. As expected, some teachers said they were not certain of this as they could not say whether changes were due to these lessons, to other subjects, or to a combination of these. Several felt they could not be sure as they had not done any tests with their students, and one answered "no", they had not observed this at all. However, the majority answered that they either thought or were certain, that

	Yes	I think so	I am unsure	No
Did the lessons improve students' problem solving?	3 27%	5 45%	2 18%	1 9%
Did the lessons have an impact on students' general learning?	2 18%	2 18%	5 45%	2 18%

Table 9.28: Number of teachers who gave each response to the questions on problem solving skills and general learning

the lessons had improved their students' problem solving skills. Teachers were also asked whether the lessons had impacted students' learning in other areas, and again many felt they could not be sure of the answer. The majority of teachers were unsure or said they had not seen an impact on students general learning. Their responses to both the problem solving and general learning questions, and the percentage of teachers this represents, are shown in table 9.28. Teachers' answers to the question on problem solving included:

- **Yes**

- “I would say yes it did... They grew in that whole problem solving aspect”
- “Yes I would say so... We do the Otago problem solving challenge and I definitely did notice at the end of last year were doing better than that then they had been doing prior to that, and I didn't teach problem solving between those tests, I taught that at the start of the year, so it wasn't that which had helped”
- “Yes... they are learning to change their thinking.”

- **I think so**

- “I probably can't say it definitely was, but I think it probably helped.”
- “We do lots of discussion on the mat... they seem to be asking questions and looking for alternative way to do things, so I think maybe it did. I haven't proven that yet. I can't be sure, but I think they were inquiring more, asking ‘does this mean this and this?’ ”
- “I hope so, it's so hard to know... I can't see how [the lessons] wouldn't help. I'm sure that they would be helpful.”

- **Unsure**

- “It’s hard to tell”

- **No**

- “Probably not. I can’t tell you a time when I noticed.”

These observations support the conclusion of the previous study, that CS, programming, and developing CT skills can improve the majority of students’ problem solving skills.

9.5.5 *Learning Transfer*

In this study, there was only one observation of learning transfer occurring for the majority of students in a class, and this was directly facilitated by the teacher. There were no observations of automatic learning transfer. The teachers who were unsure, but suspected there had been benefits in other subjects, said this was mainly for maths. They also said this was generally due to an increase in students’ confidence and engagement with maths, through the CS and programming lessons, rather than a transfer of thinking skills. One teacher believed some students had transferred some of the CT skills they learnt to other subjects, but they said this was restricted to the higher-achieving students, who already had good critical thinking skills.

The one case of the majority of students explicitly transferring their learning was in relation to reading and writing and was clearly supported by the teacher, as they focussed specifically on introducing CT skills (but not CS and programming) to these subjects. When this teacher was asked if they had observed developments in their students’ general thinking skills in other subjects, they said:

Teacher — “Yes. I think the problem solving part... when one thing doesn’t happen what do you need to do? You need to go back and check it. We say that in everything, we constantly say in writing ‘go back and check your work, check your mistakes, find the bug’ I use that now in writing. I’ll say, ‘your writing is bugged, let’s go back and find what the problem is’, things like there isn’t a comma or there isn’t a full stop. It’s the same as/when I think about what they were doing with Kid Bots it’s trial and error, and then you find the mistakes and you fix them.”

Interviewer — “And they were doing that?”

Teacher — “They were doing that, and they did it all through the rest of the year.”

Several of teachers who did not observe any cases of learning transfer said that they expected this would happen if they continued these lessons over a longer period of time, or integrated them with other subjects (after first teaching them separately). One teacher commented:

“Maybe if I had integrated the lessons a bit more it would have helped. They saw it as a completely separate topic. So I don’t think it would have had that much of an impact on their general learning.”

Comparing these results on the topic of learning transfer, to those of the previous study, indicates that learning transfer can occur, but this is unlikely to happen automatically. It is much more likely that for the majority of students to apply CT skills to other areas, this needs to be facilitated by a teacher explicitly having them apply these skills in different contexts. The differences in the amount of, and opportunities for, learning transfer between this and the previous study, supports the observations of teachers who stated that teaching these subjects over a longer time-period could contribute to students transferring CT skills.

9.5.6 Student Impacts

A prominent theme in the previous study, which emerged again in this, was how there was a small group of students in each class who was impacted particularly strongly and personally, by these lessons. One teacher described the experience of one student in their class, which was very similar to stories I had heard from several teachers in the previous study. The student they described had prior experience with Scratch programming, and became the ‘class expert’ on Scratch during programming lessons. This student was not a typical leader in the classroom, and struggled socially. They were however, highly engaged with programming, and so the teacher appointed them (along with another student) a ‘class helper’ during these lessons. The teacher described the development of this student as one of the high-points of teaching these topics:

“As a teacher there were some absolutely fabulous moments in class where the alpha male type student received help and support from Arnold⁴, because he was the resident expert, and it was new to the other student and he sat there and took directions from Arnold. But what was really lovely about that was Arnold was able to give direction and socialise, when that was the discussion point... it was one of those moments as a teacher where your heart just expands, it was one of those fabulous things.”

This teacher felt the programming lessons had been beneficial for all their other students as well.

Teachers in this study also referenced that students who were generally disengaged in class became some of the most engaged and motivated students in these lessons, and that the hands-on nature of the activities appealed to students who struggled with traditional learning methods. For example:

- “One student in particular, who has quite bad behavioural issues, really loved it and he was really really good at it... This particular student would just sort of do the bare minimum with other things, but he really got engaged with it and then he really excelled at it.”
- “There was one boy I know whose behaviour can be a little bit challenging, and he was really engaged. He really enjoyed the challenge of it. He was one of the children that really stuck out”
- Discussing impacts on particular groups of students — “I would say the ones that usually are quite tricky to engage, they enjoyed these lessons as well. The lessons were hands on, like with the parity cards, those kinds of things make a big difference.”
- “Some kids just got it who you wouldn’t think would have got it.”
- “Especially one boy who didn’t normally do a lot of work in class, he was going around asking everyone to do [the Barcode trick], and was trying to work out or find a way when it doesn’t work as well. He wanted to figure out if it maybe didn’t work every time... [The lessons] were very successful. Especially for those students who do struggle with just doing reading and writing, who find it boring sometimes. It was nice for them

⁴ name has been changed

to have something different and they really got engaged. I think every single student was engaged in the lessons.”

“I know in the past I have had kids who struggle to get their ideas onto paper so using their hands and technology, things like that, it really does help... For some kids it’s really difficult for them so this is something different.”

One quote about ‘unexpected’ students succeeding was quite different to these however. When asked about students who were particularly successful they said these students were:

“...the ones who actually have good reading skills, they loved it. They understood, and tried to understand... I thought it would maybe be just the students who are good at maths, but no. The students who are good at reading did really well. It was very interesting.”

This was very different to other teachers observations, which frequently focused on students who struggled with curriculum areas like reading and writing, or who were disengaged from learning in general. This teacher made several other references to reading skills in their interview, and emphasised that the reading skills required for the Bebras tests were too high for all but the most advanced students in their class. This observation touches on how reading skills had such a strong impact on students’ experiences with programming and the Bebras test.

One class in the study was a bilingual class, where Te Reo Māori and English were spoken, and the teacher of this class felt there was a connection between students experience working in two languages and how quickly they grasped some of the different concepts they learned. They observed:

Interviewer — “Your class is unique in the study because they are bilingual, do you have any observations on the interaction between this stuff and being a bilingual class?”

Teacher — “I think they got it really easily, but I do really believe that if they’re learning more than one language then different part of the brain is developing and I think this helped. They grasped it really easily.”

Interviewer — “do you have any idea why that might have been?”

Teacher — “I don’t really know. Because they’re constantly thinking in two different languages, the whole binary thing they just got really fast, like ‘oh yeah, it’s this

or this, it's this or this'. I think it uncomplicated things for them, but I would need more data to tell. I was really surprised at how quickly they grasped the concept."

9.6 Discussion

The results of the qualitative data collected in this study support the conclusions of the previous study. Teachers were in agreement that these lessons were highly beneficial for their students. Students practised and developed a range of CT skills, and several teachers observed improvements in their students' problem solving skills. However, based on the interviews with teachers, the amount of student skill development in this study was lower for all CT skills, and was significantly lower for four of the six CT concepts and three of the five approaches, compared with the previous study. There was also significantly less evidence of learning transfer in this study. There was only one occurrence of learning transfer, and this transfer was explicitly facilitated by the students' teacher. There are several possible explanations for these lower rates of skill development and learning transfer.

In this study all teachers were restricted to a particular set of CS Unplugged activities, very little programming took place, very little cross-curricula integration occurred, and in the previous study many participants taught over a longer time period compared to this study. These factors, which are discussed further later in this discussion, may have restricted both the students' development of CT skills and the teachers' opportunities to observe and identify these.

This also meant there were generally fewer lessons for teachers to discuss in their interviews. Instead of describing the individual lessons there was a tendency for teachers to list the lessons they conducted instead. This was considered not enough evidence to tag a section with a CT skill, as there was no description of students actually *doing* the activities. This may have contributed to the lower amount of CT tags per teacher. For some skills, there were likely fewer opportunities for these to be exercised (the most obvious one being Creating). The differences between the proportion of times each skill was tagged also could potentially be exaggerated due to the smaller sample size of this study, and that only interview data was used in this study

Despite these differences, the results of the qualitative analysis in this study support the conclusions that learning about CS, programming, and developing CT skills can have an impact on the majority of students' problem solving skills, and it is possible for learning transfer to occur, but this is unlikely to happen without teacher facilitation. CT, CS, and

programming appear to be no exception to the general consensus that learning transfer largely does not occur automatically, but through explicit teaching [3].

While these conclusions agree with those of the Open Teachers study, that students' CT and problem solving skills improved, the results of the Bebras assessment gave a different result. These conclusively showed that participants' scores on the tests did not improve more than the controls, in any case. There are several possible explanations for these results. The first of these is that despite participating in the CS Unplugged lessons and (in most cases) doing a small amount of programming, the participating students did not in fact develop CT skills. It also supports the conclusion that the lower amount of CT skill development observations was a result of students not developing these skills, rather than a lack of opportunities for teachers to observe or report development. It is also possible that there was some development of CT skills, but this was small and the Bebras challenge was not a sensitive enough assessment to identify these changes, or the Bebras challenge was not a suitable assessment tool to use.

There are several limitations to using the Bebras challenge as a CT assessment tool. These were discussed in detail in Chapter 3, section 3.4.1. One of the limitations which may have been particularly influential in this study was the challenging language used in the tests. The reading level of the Bebras questions was much higher than expected for the age groups in this study, and so students scores may have been a reflection of their reading ability, rather than their problem solving or CT skills. Reading and Maths were found to be significantly positively correlated with students' scores in all tests. As students' Reading level was also significantly positively correlated with their Maths level, this could be interpreted as Maths being the influence on students scores instead of, or alongside, Reading. However, when teachers' observations are taken into account it seems more likely that Reading was more of an influence on scores than Maths.

Teachers of both participating and control classes at each school, universally agreed that the reading ability the tests required was too high for all but the most advanced of their students. Many students asked questions during the tests which were based purely on the language used in the questions. An example of this is the first sentence of the question 'Fair share', from the 2016 Castors and Benjamins test.

"Hamid has a 4 litre beaker full of a hazardous chemical."

I discussed this question with an experienced NZ teacher, and their verdict was that just this first sentence made the question unsuitable for year 7 and 8 students, let alone

the years 4, 5, and 6's. The words 'hazardous', 'chemical', and 'beaker' are words many students would not be familiar with. They are concepts students will generally encounter for the first time when they begin High School Science. Several students were also confused by the names used in this question, Hamid and Kazim, and asked what these meant as they did not realise they were names.

The length of some of the questions which incorporated stories was also problematic. Students found the stories confusing, lost track partway through long sections of text, and could not connect these with the actual problem they were meant to be solving. While a core part of answering Bebras questions is meant to be interpreting what the questions are asking, the stories in these questions were the part that challenged students most. The questions without stories or with very short ones were much easier for students to understand. One of the recommendations given when writing Bebras questions is to incorporate a real-world context, as a way to engage students more with the questions. While this can be effective, the lengthy stories may have been more likely to disengage students and negatively impact their score. When describing the students who were strongly impacted by these topics, teachers in both this and the prior study said these were frequently the students who struggled with reading and writing. The advanced reading level of the tests may have prevented students like these from demonstrating their knowledge and CT skills, thus dragging down their scores.

There are other factors as well which may have led to students learning, or displaying, less CT. The lack of cross-curricular integration in this study may have limited students' ability to transfer the thinking skills they used during the lessons to other contexts, such as the Bebras questions. One teacher said this was likely a reason why there was no perceivable impact on students' general learning:

"Maybe if I had integrated the lessons a bit more it would have helped. They saw it as a completely separate topic. So I don't think it would have had that much of an impact on their general learning."

Another key difference between students' experiences in this study that may have impacted their skill development, was that less time was spent on programming compared to previous studies. The majority of classes spent 2-3 lessons on programming, and some did none. All of the classes which worked with Python made very little progress with the material, apart from a small number of students who had prior programming experience before the study began, or were generally high achieving at school. As discussed in Chapter 5,

programming is an effective method of reinforcing CT skills, particularly algorithmic thinking, and the lack of this likely disadvantaged students in this study compared to those in previous studies.

Furthermore, the time period (9-10 weeks, depending on when teachers started and ended the lessons) over which the study was conducted may not have been sufficient for students to practice the skills enough to apply them independently. The short time period and number of different topics teachers were asked to teach also limited their opportunities to cover concepts multiple times. Repetition of concepts was noted by teachers in previous studies as something their students needed, as they often forgot things quickly if they were not repeated. One teacher in this study said that the amount of time they spent on these topics was too short for students to consolidate their skills:

“It was too short. We would have to do it throughout the year to see the real progress.”

In future intervention studies on CS and programming developing CT skills, these are all important issues to address. A longer time period should be used when conducting a similar study to this and should allow for repetition of content. If the impact of cross-curricular integration is to be investigated time for this also needs to be allowed for, and based on the results of the Open Teachers study I recommend giving teachers plenty of freedom (with guidance) in how they choose to integrate CS and programming with other subjects. Using an age-appropriate programming environment is also crucial if students are to learn about programming.

When using the Bebras challenge as a form of CT assessment I recommend that the language of the tasks be reviewed before the test is administered. In reviewing the questions it is important that they are evaluated by an experienced teacher for the relevant age groups. As the general reading level and common vocabulary for age groups of students vary between different countries, this should be taken into account, so it would be advisable to consult a teacher with teaching experience in the country (or region) where the students are based. It should be ensured that the language used is age-appropriate, region-specific terminology is relevant and commonly known, and that questions do not involve lengthy or convoluted narratives.

9.7 Conclusion

The goal of this study was to verify conclusions from my previous work, by using an additional measurement approach, and by working with a group of teachers who were not entirely self-selected. I worked with 13 teachers at three primary schools, who taught their classes CS Unplugged lessons throughout term three of the school year. The majority of these classes also completed a small number of programming lessons. At each school, I used the Bebras Challenge to conduct pre and post-tests with both participating and non-participating classes, with the intent of measuring students' Computational Thinking skills. The results of these tests were used to investigate whether the participating students' scores improved throughout the school term. 11 of the participating teachers were interviewed about the classes and the impacts these had on their students, at the end of the school term.

The research questions I aimed to answer were similar to those of my previous studies. The first of these was: *Can learning Computer Science and programming develop, or improve, Computational Thinking skills for the majority of students?* In my previous study, I concluded that yes, CS and programming lessons can develop and improve students' CT skills. The thematic analysis of the interviews in this study agreed with this conclusion, but showed less development of these skills than the previous studies did. The results of the Bebras CT tests for each school showed no difference in CT skill development between the participating and the control students. There are several threats to the validity of these results, as discussed in section 9.6, the most prominent of which was the unsuitable vocabulary and level of reading comprehension in these challenges.

My second question was: *Can Computer Science and programming lessons, and developing Computational Thinking skills have an impact on problem solving skills, for the majority of students?* In my previous study, I had concluded that the answer to this was yes, these lessons and developing CT skills *can* improve students problem solving skills. The results of this study agree with these findings. It should be noted that this conclusion is problem solving skills *can* improve, not that they will *always* improve, as in both studies not all teachers agreed on this. Those teachers that did agree stated that this was the case for the majority of their students.

The third question for this study was: *Can Computer Science and programming lessons, and developing Computational Thinking skills have an impact on overall learning for the majority of students?* Based on the teachers' observations, learning CS, programming, and CT skills *can* have a positive impact on students' general learning in other areas. But, this is

unlikely to happen in a short period of time, and without integrating CS and programming with other subjects. This supports the conclusion of the previous study, that there can be an impact on students' general learning. The much larger amount of both learning transfer and cross-curricula integration reported in the previous study also supports the conclusion that integrating these subjects with others is a key part of facilitating learning transfer.

The fourth question was: *If these lessons do have an impact on students' Computational Thinking skills or their overall learning, do these impacts differ between students of different genders or ethnicities?* The results of the Bebras tests showed no significant difference between students' scores based on their gender or their ethnicity when the confounding factors of Reading and Maths levels were taken into account. In interviews, teachers were asked if they had observed differences between various groups of students. No teachers reported differences between the achievement or behaviour of students in their classes, based on their ethnicity. All teachers stated that both girls and boys were engaged in the lessons, and they observed no differences in their achievement. However, several teachers who discussed the particularly enthusiastic students in their classes said these were mainly (but not all) boys, who latched on intensely to the subject. While this could be seen as an indication that boys are more likely to engage with the topic, multiple teachers emphasised that despite these particularly interested male students, their students were all equally engaged. Some teachers, in this and the previous study, also remarked that this may have been to do with the more overt ways the boys displayed their interest compared to the girls, rather than a difference in their interest levels. These observations are also likely linked to the impact of gendered socialisation and match the 'magnetic-attraction' phenomena, described by Margolis and Fisher [123].

The final question I aimed to address was: *What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students?* The conclusions I came to in this study, again, validated those of my previous work. The only negative impact reported was the frustration some students experienced when they found the activities challenging, and that this behaviour was not abnormal for these students in other subjects as well. Several teachers told stories about how students they predicted would struggle with the material were unexpectedly successful, and some of the particularly interested and competent students developed leadership and social skills through helping other students.

The conclusions drawn from these results strongly support the findings of the previous studies, that:

- Computer Science and programming education *can* develop and improve the majority of students' Computational Thinking and problem solving skills.
- Computer Science, programming, and Computational Thinking education *can* have a positive impact on the majority of students' general learning, particularly when they are integrated with other subjects.
- When students develop Computational Thinking skills it *is possible* for them to transfer these skills to other learning areas, but this is unlikely to happen if students are not taught explicitly to use these skills in different contexts.

Chapter X

Conclusions

Having a foundational knowledge of computing is now a critical part of being a capable and informed citizen, due to the ubiquity of digital technologies in our society. To address this need, Computer Science, programming, and Computational Thinking have been added to school curricula around the world. As these are new subjects to school education, there are many challenges to implementing these changes. These include evaluating what computing topics need to be taught, how these can be successfully taught, and whether the introduction of these topics achieves the intended educational goals. This thesis has aimed to address these challenges, specifically in primary school curricula.

This was done by synthesising a trial curriculum, testing it in New Zealand primary schools, continually evaluating and modifying the way it was taught, and exploring the impacts it was having on students. In doing this, I worked with over 50 teachers and over two thousand primary aged students. The trial curriculum was based on the motivations for equipping students with knowledge on computing, developments in curricula around the world, literature reviews on Computational Thinking, and on teaching Computer Science and programming to young students. Resources to support people to teach these subjects were collated and developed, trialled in primary schools across New Zealand, and iteratively appraised and improved. The impacts of these lessons on students' Computational Thinking and learning were evaluated. Throughout the course of this work, resources for schools have been developed, teachers have been trained in these subjects, and findings from this research have contributed to the development of the New Zealand Digital Technologies|Hangarau Matahiko curriculum for primary schools.

10.1 Research Goals

The core goals of this research have been to confront a range of challenges in introducing Computer Science and Computational Thinking to a primary school curriculum, to investi-

gate the effects of this curriculum change on students and teachers, and to evaluate claims on the benefits of Computational Thinking for problem solving and general learning. This thesis has made significant progress towards achieving these goals.

The research questions I aimed to address, my answers to these, or where they are addressed in this thesis are listed below:

RQ1 *Can Computer Science and basic programming skills be taught in a typical primary school environment?*

Based on the results of the three studies, Computer Science and basic programming skills can be taught in a typical primary school environment.

RQ1.1 *What topics in Computer Science and programming should be trialled for teaching in this research?*

The topics that were used for teaching in this research, and why these were selected, are covered in Chapter 4, section 4.4.

RQ1.2 *What resources and training can teachers be provided with to successfully teach these?*

A range of resources for teaching Computer Science and programming to young students were discussed in Chapter 5. The specific resources and training that were provided to teachers during this research are then discussed in Chapter 6, section 6.3.

RQ1.3 *Do students learn, and how much do they learn, through these lessons?*

Based on the results of the Pilot study, the Open Teachers study, and the qualitative data obtained in the Intervention study, the majority of students did learn a significant amount about the Computer Science concepts covered by the CS Unplugged lessons. Based on the results of the Pilot study and the Open Teachers study, the majority of students were able to learn programming basics, such as selection and repetition, through using block-based programming environments.

RQ2 *Can learning Computer Science and programming develop, or improve, Computational Thinking skills for the majority of students?*

Based on the results of the Open Teachers study, and the qualitative data obtained in the Intervention study, the majority of students can develop, or improve, their Computational Thinking skills through learning Computer Science and programming.

However, based on the quantitative results of the Intervention study, students did not develop, or improve, their Computational Thinking skills through learning Computer Science and programming. Due to the limitations of the Intervention study, discussed in Chapter 9 section 9.6, and the more conclusive results of the Open Teachers study, I concluded that yes, the majority of students' can develop, or improve their Computational Thinking skills through learning Computer Science and programming.

RQ2.1 *How can students' Computational Thinking skills, and changes in these, be measured?*

The potential methods for assessing students' Computational Thinking skills, and the ones I chose to use, are discussed in Chapter 3 section 3.4.1, and Chapter 6 section 6.4.

RQ2.2 *Do students exercise or employ Computational Thinking skills during Computer Science and programming lessons, based on teacher observations?*

Based on teacher observations throughout each of the three studies, the majority of students do exercise and employ Computational Thinking skills during Computer Science and programming lessons.

RQ2.3 *Is there evidence of students' Computational Thinking skills improving, as a result of the Computer Science and programming lessons, based on objective assessment and teacher observations?*

Based on the objective assessment carried out in the Intervention study, there is no evidence of students' Computational Thinking skills improving. Based on teacher observations throughout each of the three studies however, there is evidence of the majority of students' Computational Thinking skills improving.

RQ3 *Can learning Computer Science and programming, and developing Computational Thinking skills, have an impact on problem solving skills or general learning, for the majority of students?*

Based on the results of the three studies, learning Computer Science and programming, and developing Computational Thinking skills can have a positive impact on problem solving skills and general learning for the majority of students.

RQ3.1 *Do students exercise or employ general problem solving skills during Computer Science and programming lessons, based on teacher observations?*

Based on teacher observations throughout each of the three studies, the majority of students do exercise and employ general problem solving skills during Computer Science and programming lessons.

RQ3.2 *Is there evidence of students' problem solving skills improving, based on teacher observations?*

Based on teacher observations throughout each of the three studies, the majority of students' problem solving skills did improve.

RQ3.3 *Is there evidence of the Computer Science and programming lessons having impacts on students' learning in other curriculum areas, based on teacher observations?*

Based on teacher observations throughout each of the three studies, the Computer Science and programming lessons did not have any negative impacts on students' learning in other curriculum areas, and for the majority of students, the lessons had positive impacts on their learning in other curriculum areas.

RQ3.4 *Is there evidence that developing Computational Thinking skills results in these skills being transferred to other areas of students' learning, based on teacher observations?*

Based on teacher observations throughout each of the three studies, the vast majority of students did not transfer their Computational Thinking skills to other learning areas. A minority of teachers did observe their students transferring their skills to other areas of learning, showing that this is possible, but this occurred either as a result of cross-curricula integration or by explicitly teaching students to apply these skills in different contexts.

RQ4 *What other impacts can learning Computer Science and programming, and developing Computational Thinking skills have on students?*

Based on the results of the Pilot study, the Open Teachers study, and teachers' observations in the Intervention study, there were many positive impacts of these lessons on students, both academically and socially. These are detailed in the results and discussion sections of Chapters 7, 8, and 9.

10.2 Contributions

This thesis has made seven primary contributions to knowledge in the domain of Computer Science and Computational Thinking education. It has:

1. Provided evidence to support the claim that Computational Thinking can benefit the majority of students' problem solving skills, in both computing and unrelated fields. This claim had not been tested or validated prior to this research.
2. Provided evidence against the claim that by learning Computational Thinking skills through Computer Science and programming students will automatically transfer this learning to other areas. However, there is some evidence that the majority of students can transfer these skills *if* they are explicitly taught to apply Computational Thinking skills in different contexts, or if Computer Science and programming are taught in a cross-curricula way.
3. Mapped the general skills (the Key Competencies, such as working together, social skills, and resilience) that are the core principles the New Zealand Curriculum is based on, to Computer Science, programming, and Computational Thinking.
4. Shown that concepts from Computer Science and programming can be taught in a suitable and successful way for all primary school age groups (ages 5 - 13 years old). The majority of prior research has focused on tertiary level education or selected students, whereas the results here apply to typical primary school classroom environments, with generalist teachers; and for students with a wide array of ability levels, subject interest, and difficulties or strengths with learning, with the caveat that teachers are adequately supported and resourced.
5. Shown that not only could these subjects be taught to students with a wide range of abilities, these students were all (apart from students with severe developmental disabilities) able to achieve success to some degree in these classes, and some students who struggled with learning or were seen as low achieving were particularly successful, often to the surprise of the teacher. Programming and topics from Computer Science appeared to appeal strongly to groups of students who have not been successful within the existing education system, re-engaged them with learning, provided them with

new ways to learn and demonstrate their learning which are more suitable for them, and have facilitated the development of other skills, particularly social skills, for these students.

6. Given an evaluation of multiple Computer Science Unplugged activities, which was done in conjunction with participating teachers. The Computer Science Unplugged resources were found to be easy for teachers to use, the activities were engaging and enjoyable for students, and teachers felt that both they and their students learnt from these. The use of these activities led teachers to continue teaching Computer Science topics after the conclusion of the study; developed their interest in, and knowledge of the subjects; and enthused many to encourage and support other teachers to integrate Computer Science and programming into their own teaching.
7. Collated information on the design of Computer Science, programming, and Computational Thinking curricula, and a given a framework for a curriculum, parts of which have led directly into the design of the new New Zealand Digital Technologies|Hangarau Matahiko curriculum.

10.3 Limitations

As was discussed in Chapter 6, section 6.5, throughout my work the most influential limitation was that my involvement in the NZ curriculum redevelopment, and in providing support for the implementation of this curriculum, took priority over the process of my research. This, and the pragmatic research approach this required, limited the repeatability and rigour of the studies conducted. As the curriculum changed throughout my research there were frequent changes in the content I taught and asked teachers to teach, the training and resources our research group was providing, and the ability of teachers to participate in the research. These outside influences introduced a huge number of uncontrolled variables to the studies.

There are also several threats to the validity of this research that must be acknowledged, which were also discussed in Chapter 6, section 6.5. While the Bebras challenge has many merits, its suitability as an assessment tool for CT is not certain. Through the interview process and analysis, there were risks of bias being introduced. These were the potential for acquiescence bias from participants, and subjectivity in my analysis of these interviews. As there was no second person who examined the interview transcripts, it was also not possible

for my conclusions to be corroborated. Relying on teachers' impressions and opinions of their students' learning also has limitations, as it is not possible for a teacher to have full access to the extent of a student's knowledge this way.

10.4 Future work

There is an extensive range of opportunities for future work in this area, particularly in the field of Computational Thinking assessment. As explored in chapters 3 and 9, reliable and objective means of measuring students skills in this area are not widely available. While observational information is extremely useful, relying on subjective information only is limiting. Having other forms of Computational Thinking assessment is necessary if it, Computer Science, and programming, are to become established curriculum subjects. Achievement measurements are also crucial for improving the teaching of these subjects and making changes to curriculum content when necessary.

Future work could also build on the findings of this thesis through longitudinal studies, to examine how students' skills and knowledge develop over time. Several teachers involved in this work voiced concerns about whether there was going to be a clear progression of lessons and content at each school year level. They said that if there was not new material for their students to learn when they moved on to their next year group, they would quickly lose interest and would not continue learning.

Observing students over more than one year would also be an opportunity to further investigate how this subject material impacts specific groups of students. In particular, how this material greatly benefited students who struggled with the traditional learning methods of reading and writing, were disengaged from learning, and who found it difficult to communicate their learning and knowledge. The many cases of teachers describing how these students became engaged with not just this material, but also other subjects; found they could convey their learning more successfully through programming; and in several cases went through a great deal of personal development, would be a compelling phenomenon to investigate in future research.

Appendix I

Tables and Figures

A.1 Curriculum comparisons

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
England Key stage 1 (students from 5 to 7 years old)					
Understand what algorithms are	Create and debug simple programs; relationship with algorithms			Create, organise, store, manipulate and retrieve content	Recognise use beyond school; use safely and respect
Australia F-2 (students from 5 to 8 years old)					
Sorting and patterns	Step-by-step procedures; sequenced instructions; robotic toys	Patterns and symbols; pixels and file size	Hardware and software components; features of a device; data transfer	Capture and manipulate data e.g. photo; downloading information; collect, organise, present data; use common software	Ergonomics, digital devices in everyday life; collaboration; ethical and safe use
CSTA L1:3 (students from 5 to 8 years old)					
Logical problems	Purpose of software; turtle instructions	How 0s and 1s represent information	Use input/output devices	Writing tools; organising information (e.g. sorting); gather and communicate information; age appropriate research; use and create multimedia; concept mapping	Work collaboratively; careers that use computing; legal and ethical behaviour

Table A.1: Comparison of computing curricula for early primary school years [67]

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
England Key stage 2 (students from 7 to 11 years old)					
Decompose problems; explain and correct errors in algorithms	Sequence, selection and repetition; variables; input/output; design, write and debug for specific goals		Understand networks	Select, use and combine software to work with data; use search technology; create content	Use safely, respectfully and responsibly; appropriate behaviour; reporting concerns
Australia 3-4 (students from 8 to 10 years old)					
Steps to solve simple problems; sequences of steps	Interactive application in visual programming language (sequence and branching; input)	Representation of text, image, sound etc. as data; codes and symbols	Peripheral devices; data capture; data transfer	Collect, organise and present data; simple spreadsheets and charts	How systems meet community and personal needs; evaluate adequacy of a solution; collaboratively plan creation and communication of information; ethical decisions and behaviour; social media
CSTA L1:6 (students from 8 to 12 years old)					
Algorithmic problem solving, basic algorithms e.g. search, sort (computer-free); decomposition	Step-by-step program; block based visual programming	Representing alphanumeric data as strings of bits	Identify simple hardware and software problems; information is from networks;	Simulation; productivity tools; online resources; tools for problem solving and learning; Multimedia collaborative authoring; gather and manipulate data; access information, navigate web, intelligent systems	Connection of CS with other fields; collaborative problem solving and innovation; jobs that require computing; keyboard proficiency; pervasiveness of computing; human/machine distinction; appropriate use; impact of technology; reliability of information; ethical issues

Table A.2: Comparison of computing curricula for middle primary school years [67]

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
Australia 5-6 (students from 10 to 12 years old)					
Design/modify common algorithms e.g. sorting; multiple ways to represent instructions	Branching, iteration, user input; visual programming language	All data is binary; small binary numbers; patterns in binary	Main components and functions; how data is transmitted; cloud; networks; emerging systems	Collect, validate real world data; spreadsheet calculations; data visualisation; evaluate and design solutions	Design/evaluate a user interface; sustainability; design for need; ethics in networks; energy saving; online collaboration and communication; safe practices
England Key stage 3 (students from 11 to 14 years old)					
Understand and compare key algorithms e.g. searching and sorting	2 or more languages, one textual; composite data types; functions/procedures	Boolean logic; binary number operations; binary representation of instructions, data types	Hardware and software components	Creative projects with multiple applications; analysing data; model real world	Work with artefacts with attention to trustworthiness, design and usability; protecting identity and privacy; appropriate use; reporting concerns
CSTA Level 2 (students from 11 to 15 years old)					
Algorithmic problem solving; parallelization; evaluate different algorithms for same problem; act out searching/sorting algorithms; abstraction to decompose problems; practical application of algorithms	Implement with looping, conditionals, logic, expressions, variables and functions	Represent text, sounds, pictures, numbers	Hierarchy in computing; executing programs; hardware/software relationship; terminology; troubleshooting; major components	Charts and diagrams; simulation for problem solving; model accuracy; productivity/multimedia tools; select tools; design and publish (including web); collect and analyse data; accuracy of information	Connection with maths; interdisciplinary Computational Thinking; collaborative work; good security practice; interdisciplinary careers; dispositions for collaborative work; human vs machine intelligence; ethical and legal behaviour; impact on society; ethical issues; global equity
Australia 7-8 (students from 12 to 14 years old)					
Decomposition; design algorithms to search, sort, merge etc.; check algorithm; pseudocode and diagrams	Develop and modify programs; digital game; robot programming; programs with controlled repetition	Text encoding; bitmap and vector images; RGB colour; binary conversion	Network components; reliability and speed; protocols; cloud vs client based systems	Advanced use of search engine; acquiring data; authenticity of data; visualise and filter data; database queries; model of real world; create web-based project	Designing for users; environmental considerations; interface design; evaluate system suitability; e-waste; online behaviour; project management

Table A.3: Comparison of computing curricula for late primary school years [67]

A.2 Proposed computing curricula for NZ

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
NZC level 1, year 1–3 (students from 5 to 7 years old)					
Understand what algorithms are and follow an algorithm	Step-by-step procedures; sequenced instructions; turtle instructions (e.g. robotic toys)	(none)	Use common input/output devices (e.g. keyboard, pointing device, touch screen)	Create, organise, store, manipulate and retrieve content	Use technology safely and respectfully, keeping personal information private; identify where to go for help and support when they have concerns about content or contact on the internet or other online technologies; recognise use beyond school
NZC level 2, year 3–5 (students from 7 to 9 years old)					
Decompose problems into steps; explain and correct errors in algorithms	Block-based/visual programming including simple iteration	How two different symbols can represent information (e.g. binary numbers)	Peripheral devices; data capture; data transfer	Select, use and combine software to work with data; use search technology; create content	Use safely, respectfully and responsibly; appropriate behaviour; reporting concerns
NZC level 3, year 5–8 (students from 9 to 12 years old)					
Recognise that there can be multiple algorithms for the same problem (e.g. searching, sorting)	Block-based/visual programming including selection and iteration, user input, output, variables	Simple binary representations: integers, text; file types (extensions) and sizes	Identify simple hardware and software problems; understand networks	Create, organise, manipulate and retrieve content	Safe/ethical use of digital devices; impact of computers on humans

Table A.4: Proposed computing curricula for NZ levels 1-3, conservative version

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
NZC level 4, year 8–9 (students from 12 to 13 years old)					
Informally compare multiple algorithms for the same problem (e.g. searching, sorting); design simple algorithms to solve problems for non-computing situations (e.g. finding duplicates in a list, merging, simple turtle movements)	Using a block-based or text-based language: Recognising data types (e.g. numbers/integers, text/strings); simple Boolean expressions (for use in 'if' statements)	Representations of integers, text, images, sound	Main components and functions; networks and the cloud; emerging systems	Creative projects with multiple applications; analysing data; model real world	Ethical decisions and behaviour; social networks; careers in computing
NZC level 5, year 10 (students from 14 to 15 years old)					
Evaluate multiple algorithms for the same problem (e.g. searching, sorting), recognising best/worst cases; Recognise the problems with exponential time algorithms; Design (plan) simple algorithms with pseudo-code or diagrams before writing a program.	Using a block-based or text-based language: using simple modules (functions/procedures); simple linear data types (lists/arrays); simple text-based programming in comparison with a block-based language	Compression of data (simple methods); comparison of representations (e.g. compressed and uncompressed, ASCII and Unicode)	Simple troubleshooting (devices and network)	Developing multimedia/web products; simulation	Careers in computing; create, revise and re-purpose digital artefacts with attention to trustworthiness, design and usability; protecting identity and privacy; appropriate use; reporting concerns

Table A.5: Proposed computing curricula for NZ levels 4-5, conservative version.

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
NZC level 1, year 1–3 (students from 5 to 7 years old)					
Understand what algorithms are and follow an algorithm; sorting and patterns	Create and debug simple programs (e.g. turtle instructions) with simple sequencing and repetition	How 0s and 1s represent information; patterns and symbols; pixels and file size	Describe hardware and software components; input/output devices	Create, organise, store, manipulate and retrieve content, including multi-media	Ergonomics, digital devices in everyday life; ethical and safe use
NZC level 2, year 3–5 (students from 7 to 9 years old)					
Steps to solve a simple problem	Design, implement, test and debug an interactive application in a visual programming language (sequence and selection; input)	Representation of text and images using binary; codes and symbols	Peripheral devices; data capture; data transfer; identify simple software problems	Select, use and combine software to collect, organise and present data; simple spreadsheets and charts	How systems meet community and personal needs; evaluate adequacy of a solution; collaboratively plan creation and communication of information; ethical decisions and behaviour; social media
NZC level 3, year 5–7 (students from 9 to 12 years old)					
Algorithmic problem solving; basic algorithms e.g. search, sort (computer-free); decomposition; explain and correct errors in algorithms	Programming in a visual or textual language to meet a given requirement (using sequencing, selection and iteration; including Boolean expressions with more than one operator)	Representation of text, image, and sound using binary	How data is transmitted; configuring and troubleshooting simple systems	Creative projects with multiple applications; analysing data; model real world	Careers that use computing; design/evaluate a user interface; sustainability; design for need; ethics in networks; energy saving; online collaboration and communication; safe practices

Table A.6: Proposed computing curricula for NZ levels 1-3, ambitious version.

Algorithms	Programming	Data representation	Digital devices and infrastructure	Digital applications	Humans and computers
NZC level 4, year 8–9 (students from 12 to 13 years old)					
Performance of an algorithm for an input of size n (e.g. searching, sorting)	Two or more languages, one textual; simple linear data types (lists/arrays); functions/procedures	All data is binary; patterns in binary; quality vs. file size for text, images, sound.	Hardware/software relationship; simple electronics; terminology; troubleshooting	Advanced search engine use; authenticity of data; visualise and filter data; database queries; simulation; web-based project	Create, revise and re-purpose digital artefacts with attention to trustworthiness, design and usability; protecting identity and privacy; appropriate use; reporting concerns ethical issues; global equity; sustainability and e-waste evaluation
NZC level 5 (students from 14 to 15 years old)					
Relative performance of two algorithms for an input of size n (e.g. searching, sorting); difference between an algorithm and a program	Implement with looping, conditionals, logic, expressions, variables and functions in a text-based language	Binary number operations; Boolean logic; binary representation of instructions; introductory coding (e.g. compression)	Network components; reliability and speed; protocols; cloud vs client based systems; security; digital electronics (simple embedded systems)	Presenting data from database; analyse and visualise data; model processes/simulation; on-line interactive solutions (web)	Human vs machine intelligence; basic HCI concepts; ethical and legal behaviour; impact on society; evaluate system suitability; basic software engineering concepts

Table A.7: Proposed computing curricula for NZ levels 4-5, ambitious version.

A.3 Pilot Study

A.3.1 Bebras challenge and quiz results

Year level	N	Mean	Minimum	Maximum
Year 7	127	40.41% ($\sigma=15.05\%$)	9.44%	84.44%
Year 8	98	39.95% ($\sigma=19.78\%$)	4.44%	100%
Total	225	40.21% ($\sigma=17.23\%$)	4.44%	100%

Table A.8: Bebras score percent across year 7 and 8 students

Gender	N	Mean	Minimum	Maximum
Female	115	40.00% ($\sigma=16.16\%$)	8.89%	100%
Male	110	40.43% ($\sigma=18.36\%$)	4.44%	93.33%
Total	225	40.21% ($\sigma=17.23\%$)	4.44%	100%

Table A.9: Bebras score percent across female and male students

Ethnicity	N	Mean	Minimum	Maximum
Māori/ Pacifica	27	40.21% ($\sigma=16.70\%$)	8.89%	77.78%
Other	108	41.66% ($\sigma=17.91\%$)	4.44%	100%
Total	135	41.37% ($\sigma=17.62\%$)	4.44%	100%

Table A.10: Bebras score percent across Māori/Pacifica, and non Māori/Pacifica students

Year level	N	Mean	Minimum	Maximum
Year 7	239	66.25% ($\sigma=16.61\%$)	26.67%	93.33%
Year 8	127	56.12% ($\sigma=18.83\%$)	13.33%	100%
Total	366	62.73% ($\sigma=18.05\%$)	13.33%	100%

Table A.11: Data representation quiz percent across year 7 and 8 students

Gender	N	Mean	Minimum	Maximum
Female	167	62.95% ($\sigma=16.58\%$)	26.67%	93.33%
Male	160	63.25% ($\sigma=19.64\%$)	13.33%	100%
Total	327	63.10% ($\sigma=18.12\%$)	13.33%	100%

Table A.12: Data representation quiz percent across female and male students

Ethnicity	N	Mean	Minimum	Maximum
Māori/ Pacifica	60	61.89% ($\sigma=16.97\%$)	26.67%	100%
Other	227	63.94% ($\sigma=18.03\%$)	13.33%	100%
Total	287	63.51% ($\sigma=17.80\%$)	13.33%	100%

Table A.13: Data representation percent across Māori/Pacifica, and non Māori/Pacifica students

Year level	N	Mean	Minimum	Maximum
Year 7	238	61.68% ($\sigma=16.98\%$)	10%	90%
Year 8	137	59.12% ($\sigma=18.73\%$)	10%	100%
Total	375	60.75% ($\sigma=17.66\%$)	10%	100%

Table A.14: Programming quiz score across year 7 and year 8 students

Gender	N	Mean	Minimum	Maximum
Female	179	61.90% ($\sigma=17.18\%$)	20%	90%
Male	154	61.10% ($\sigma=17.58\%$)	10%	100%
Total	333	61.53% ($\sigma=17.34\%$)	10%	100%

Table A.15: Programming quiz score across female and male students

Ethnicity	N	Mean	Minimum	Maximum
Māori/ Pacifica	65	58.77% ($\sigma=13.52\%$)	30%	90%
Other	237	62.49% ($\sigma=17.93\%$)	10%	100%
Total	302	61.69% ($\sigma=17.12\%$)	10%	100%

Table A.16: Programming quiz score across Māori/Pacifica, and non Māori/Pacifica students

A.3.2 Attitude survey results

		Q2	Q3	Q4	Q5	Q6	Q7
Q1) Programming enjoyment	Girls	0.57	0.55	0.41	0.28	0.50	0.29
	Boys	0.43	0.39	0.35	0.26	0.51	0.34
Q2) Data-rep enjoyment	Girls		0.50	0.52	0.49	0.37	0.47
	Boys		0.41	0.36	0.32	0.39	0.58
Q3) Continue learning	Girls	0.50		0.56	0.37	0.37	0.40
	Boys	0.41		0.56	0.38	0.36	0.37
Q4) Consider a career	Girls	0.52	0.56		0.64	0.34	0.38
	Boys	0.36	0.56		0.66	0.43	0.34
Q5) Prior career interest	Girls	0.45	0.37	0.64		0.26	0.26
	Boys	0.32	0.37	0.66		0.31	0.30
Q6) Programming confidence	Girls	0.37	0.37	0.34	0.26		0.29
	Boys	0.39	0.36	0.43	0.31		0.51
Q7) Data-rep confidence	Girls	0.47	0.40	0.38	0.26	0.29	
	Boys	0.58	0.40	0.34	0.30	0.51	

All correlations significant at the $p < 0.001$ level.

Table A.17: Correlations between all students' answers to the attitude survey questions, split by gender.

		Q2	Q3	Q4	Q5	Q6	Q7
Q1) Programming enjoyment	Year 7	0.48	0.46	0.35	0.23	0.51	0.34
	Year 8	0.55	0.53	0.48	0.37	0.55	0.33
Q2) Data-rep enjoyment	Year 7		0.43	0.46	0.38	0.45	0.52
	Year 8		0.53	0.45	0.40	0.37	0.57
Q3) Continue learning	Year 7	0.43		0.62	0.40	0.35	0.41
	Year 8	0.53		0.60	0.45	0.44	0.39
Q4) Consider a career	Year 7	0.46	0.62		0.71	0.36	0.40
	Year 8	0.45	0.60		0.61	0.48	0.37
Q5) Prior career interest	Year 7	0.38	0.40	0.71		0.28	0.32
	Year 8	0.40	0.45	0.60		0.36	0.28
Q6) Programming confidence	Year 7	0.45	0.35	0.36	0.28		0.50
	Year 8	0.37	0.44	0.48	0.36		0.39
Q7) Data-rep confidence	Year 7	0.52	0.41	0.40	0.32	0.50	
	Year 8	0.57	0.39	0.37	0.28	0.39	

All correlations significant at the $p < 0.001$ level.

Table A.18: Correlations between students' answers to the attitude survey questions, split by year group.

	Q2	Q3	Q4	Q5	Q6	Q7
Q1) Programming enjoyment	0.61	0.49	0.38	0.17*	0.48	0.43
Q2) Data-rep enjoyment		0.51	0.43	0.35	0.52	0.60
Q3) Continue learning	0.51		0.61	0.37	0.36	0.41
Q4) Consider a career	0.43	0.61		0.62	0.40	0.35
Q5) Prior career interest	0.35	0.37	0.62		0.30	0.29
Q6) Programming confidence	0.52	0.36	0.40	0.30		0.44
Q7) Data-rep confidence	0.60	0.41	0.35	0.29	0.44	

* $p > 0.05$ ($p = 0.09$), all other correlations significant at the $p < 0.001$ level.

Table A.19: Correlations between Māori and Pacifica students' answers to the attitude survey questions.

Female vs male students		
Question	Effect size	Significance
Q1) Programming enjoyment	0.35	p<0.001
Q2) Data-rep enjoyment	0.30	p<0.001
Q3) Continue learning	0.51	p<0.001
Q4) Consider career	0.58	p<0.001
Q5) Previously considered career	0.51	p<0.001
Q6) Programming confidence	0.32	p<0.001
Q7) Data-rep confidence	0.27	p<0.001

Table A.20: Differences between female and male students' answers to the attitude survey

Year 7, female vs male students		
Question	Effect size	Significance
Q1) Programming enjoyment	0.34	p<0.01
Q2) Data-rep enjoyment	0.25	$p>0.05$
Q3) Continue learning	0.42	p<0.01
Q4) Consider career	0.49	p<0.001
Q5) Previously considered career	0.38	p<0.01
Q6) Programming confidence	0.41	p<0.01
Q7) Data-rep confidence	0.35	p<0.01

Table A.21: Differences between Year 7 female and male students' answers to the attitude survey

Year 8, female vs male students		
Question	Effect size	Significance
Q1) Programming enjoyment	0.37	p<0.01
Q2) Data-rep enjoyment	0.37	p<0.01
Q3) Continue learning	0.59	p<0.001
Q4) Consider career	0.67	p<0.001
Q5) Previously considered career	0.65	p<0.001
Q6) Programming confidence	0.25	p>0.05
Q7) Data-rep confidence	0.20	p>0.05

Table A.22: Differences between Year 8 female and male students' answers to the attitude survey

Year 7 female vs year 8 female students		
Question	Effect size	Significance
Q1) Programming enjoyment	0.31	p<0.05
Q2) Data-rep enjoyment	0.10	p>0.05
Q3) Continue learning	0.30	p<0.05
Q4) Consider career	0.09	p>0.05
Q5) Previously considered career	0.03	p>0.05
Q6) Programming confidence	0.07	p>0.05
Q7) Data-rep confidence	0.06	p>0.05

Table A.23: Differences between female Year 7 students' and female Year 8 students' answers to the attitude survey

Year 7 male vs year 8 male students		
Question	Effect size	Significance
Q1) Programming enjoyment	0.21	p>0.05
Q2) Data-rep enjoyment	0.03	p>0.05
Q3) Continue learning	0.11	p>0.05
Q4) Consider career	0.22	p>0.05
Q5) Previously considered career	0.22	p>0.05
Q6) Programming confidence	0.21	p>0.05
Q7) Data-rep confidence	0.18	p>0.05

Table A.24: Differences between male Year 7 students' and male Year 8 students' answers to the attitude survey

Binary Score		
Question	Pearson corr.	Sig.
Q1) Programming enjoyment	0.139	p<0.05
Q2) Data-rep enjoyment	0.104	p=0.08
Q3) Continue learning	0.09	p=0.13
Q4) Consider career	0.03	p=0.58
Q5) Previously considered career	0.01	p=0.87
Q6) Programming confidence	0.11	p=0.07
Q7) Data-rep confidence	0.18	p<0.05

Table A.25: Pearson correlations between student's score in the Data-rep quiz and their survey answers

Programming Score		
Question	Pearson corr.	Sig.
Q1) Programming enjoyment	0.19	p<0.05
Q2) Data-rep enjoyment	0.08	p=0.16
Q3) Continue learning	0.13	p<0.05
Q4) Consider career	0.58	p=0.10
Q5) Previously considered career	-0.01	p=0.861
Q6) Programming confidence	0.16	p<0.05
Q7) Data-rep confidence	0.18	p<0.05

Table A.26: Pearson correlations between student's score in the Programming quiz and their survey answers

	Binary total		Prog total	
	Pearson corr.	Sig. (2-tailed)	Pearson corr.	Sig. (2-tailed)
Female students				
Binary confidence	0.028	p>0.05	0.103	p>0.05
Prog confidence	0.139	p>0.05	0.090	p>0.05
Male students				
Binary confidence	0.300	p<0.01	0.252	p<0.01
Prog confidence	0.079	p>0.05	0.248	p<0.01

Table A.27: Pearson correlations between student's gender and their confidence

	Binary total		Prog total	
Māori/Pacifica students	Pearson corr.	Sig. (2-tailed)	Pearson corr.	Sig. (2-tailed)
Binary confidence	0.01	p>0.05	0.07	p>0.05
Prog confidence	0.04	p>0.05	-0.16	p>0.05
Non-Māori/Pacifica students				
Binary confidence	0.20	p<0.01	0.21	p<0.01
Prog confidence	0.10	p>0.05	0.22	p<0.01

Table A.28: Pearson correlations between student's ethnicity and their confidence

Appendix II

Pilot Study Quizzes

B.1 Data Representation Quiz

Section 1.

1. What code do computers use to store information?

- a) Java b) Binary c) Python d) HTML

Answer: b) Binary

2. What are the two digits used in binary code?

- a) 10, 1 b) 1, 1 c) 0, 1 d) 0, 0

Answer: c) 0, 1

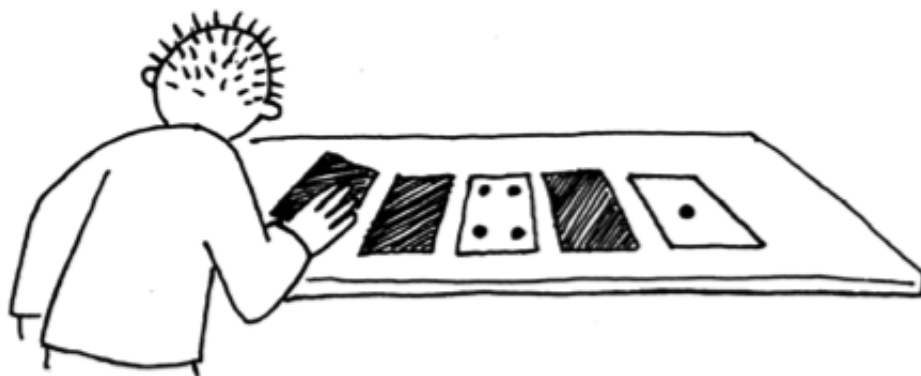
3. Everything that appears on your screen is merely represented as a series of “0’s” and “1’s” inside the computer, including number, punctuation, pictures, and colours.

- a) True b) False

Answer: a) True

Section 2.

Use the image below to help you answer the next several questions. We are only using these five cards today.



4. Using binary, is there more than one way to make any number?

- a) Yes b) No

Answer: b) No

5. What is the biggest number you can make with these 5 cards?

- a) 16 b) 30 c) 31 d) 33

Answer: c) 31

6. What is the smallest number you can make with these 5 cards?

- a) 1 b) 5 c) 0 d) 31

Answer: c) 0

7. Is there any whole number you can't make, between the biggest and the smallest number?

- a) Yes b) No

Answer: b) No

8. To make the number 19 requires the cards representing 16, 2, and 1.

- a) True b) False

Answer: a) True

9. In binary 10101 represents which of the following numbers?

- a) 17 b) 23 c) 21 d) 3

Answer: c) 21

10. If you add the numbers up from the beginning the sum will always be one less than the next number in the sequence ¹

- a) True b) False

Answer: a) True

11. When you put a zero on the right hand side of a binary number, the number doubles.

- a) True b) False

Answer: a) True

Section 3.

Use the table below to help you answer the next several questions.

1	2	3	4	5	6	7	8	9	10	11	12	13
a	b	c	d	e	f	g	h	i	j	k	l	m
14	15	16	17	18	19	20	21	22	23	24	25	26
n	o	p	q	r	s	t	u	v	w	x	y	z

12. Read the sequence of binary using the table above to help you work out what my favourite cartoon character is. 11001, 01111, 00111, 01001, 00000, 00010, 00101, 00001, 10010. Note: Don't forget 00000 represents a space between words.

- a) betty boop b) daffy duck c) yogi bear d) bugs bunny

Answer: c) yogi bear

¹ This wording is not very clear or technically correct. However it matched the wording the teacher had used in class when teaching this particular concept.

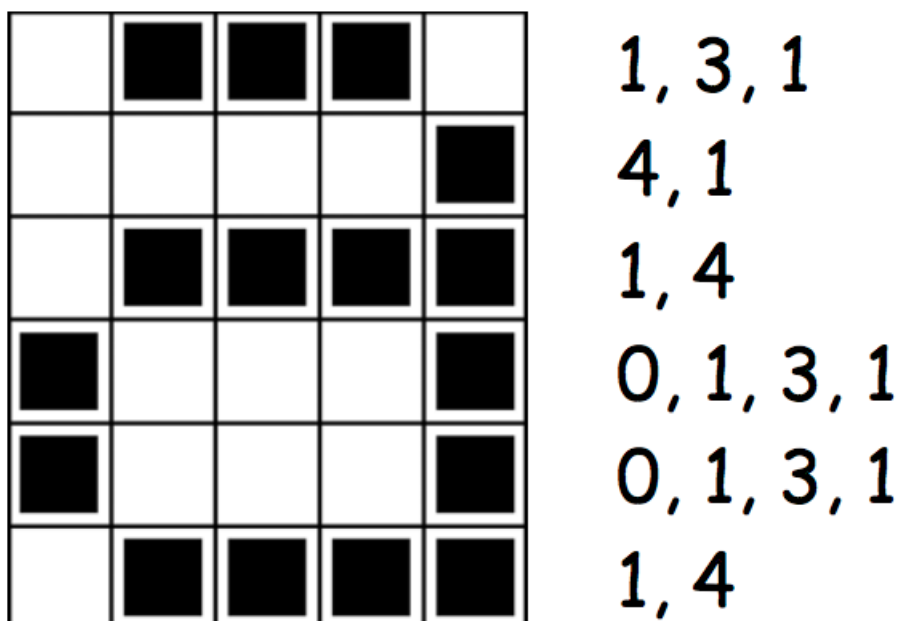
13. Read the sequence of binary using the table above to help you. If I were an animal I would be a? 01000, 01001, 10000, 10000, 01111, 10000, 01111, 10100, 00001, 01101, 10101, 10011

- a) rhinoceros b) komodo dragon c) chinchilla d) hippopotamus

Answer: d) hippopotamus

Section 4.

Computers store drawings, photographs and other pictures using only numbers. Computer screens are divided up into a grid of small dots called pixels (picture elements).



14. In a black and white picture, each pixel is either black or white. The first number always relates to the number of black pixels. Use the picture above to help you answer the question².

- a) True b) False

Answer: b) False

² Pictures on computers are not actually represented in this same way. This is the way the concept was taught in the CS Unplugged activities however

15. If the first pixel in the row is black, the line will begin with a 1.

- a) True b) False

Answer: b) False

B.2 Scratch Programming Quiz

1. An input is any information provided to the program. Which of the following listed below in the most commonly used input device?

- a) Mouse b) Camera c) Keyboard d) Microphone

Answer: a) Mouse, or c) Keyboard

2. Output is any information (or effect) that a program produces: Select from below the best answer. a) Sound is an example of an output.

b) Text is an example of an output.

c) Pictures are an example of an output.

d) All of the above are examples of outputs.

Answer: d) All of the above are examples of outputs.

3. Which of the following best defines the concept of what we mean by a Scratch block?

a) A command to one of the sprites.

b) A block on the stage.

c) A sprite on the stage.

d) When one sprite interferes with another.

Answer: a) A command to one of the sprites.

4. How is a sequence of commands created in Scratch?

a) By linking blocks together with connector lines.

b) By dragging and snapping blocks together.

c) By simply dragging and dropping blocks onto the script area.

d) By surrounding blocks with a sequence block.

Answer: b) By dragging and snapping blocks together, or c) By simply dragging and dropping blocks onto the script area.

5. The process of writing the computer instructions is called?

a) Compiling b) Interpreting c) Coding d) Debugging

Answer: c) Coding

6. Refer to the below program to answer this question:



Which shape will be created from running the above program?

a) b) c) d)

Answer: c) Note: this answer is dependant on the sprite facing towards the right of the screen when the program is first run. This is the default direction a Scratch sprite faces in when it is first added to a program.

7. An input is any information provided to the program.

a) True b) False

Answer: a) True

8. Programming requires three primary skills, which item below is not one of the primary skills you need to master?

- a) selection b) reception c) repetition d) sequence

Answer: b) reception

9. What is a variable?

- a) A long-term storage of songs, photos and documents.
- b) Short-term storage of numeric values.
- c) A loop that runs at varying speeds.
- d) A memory area that holds a single value.

Answer: d) A memory area that holds a single value.

10. When thinking about variables for your program, the rule of thumb we used is to create one variable for each:

- a) Sprite in your program.
- b) Repetition block in your program.
- c) For each value you need to keep track of.
- d) For each player in your game.

Answer: c) For each value you need to keep track of.

Appendix III

Ethics approval documents

Contents:

- Approval letter for the Pilot Study.
- 2015 approval letter for the Open Teachers Study.
- 2016 approval letter for the Open Teachers Study
- Approval letter for the 2017 computational thinking skills study.

HUMAN ETHICS COMMITTEE

Secretary, Lynda Griffioen
Email: human-ethics@canterbury.ac.nz

Ref: HEC 2014/143

19 November 2014

Caitlin Duncan
Department of Computer Science & Software Engineering
UNIVERSITY OF CANTERBURY

Dear Caitlin

The Human Ethics Committee advises that your research proposal “Student attitudes towards computer science and programming curriculum” has been considered and approved.

Please note that this approval is subject to the incorporation of the amendments you have provided in your emails of 12 and 15 November 2014.

Best wishes for your project.

Yours sincerely



Lindsey MacDonald
Chair
University of Canterbury Human Ethics Committee

HUMAN ETHICS COMMITTEE

Secretary, Lynda Griffioen
Email: human-ethics@canterbury.ac.nz

Ref: 2015/33/ERHEC

5 October 2015

Caitlin Duncan
Department of Computer Science & Software Engineering
UNIVERSITY OF CANTERBURY

Dear Caitlin

Thank you for providing the revised documents in support of your application to the Educational Research Human Ethics Committee. I am very pleased to inform you that your research proposal "Computer science and programming resources for New Zealand primary school teachers" has been granted ethical approval.

Please note that this approval is subject to the incorporation of the amendments you have provided in your email of 29 September 2015.

Should circumstances relevant to this current application change you are required to reapply for ethical approval.

If you have any questions regarding this approval, please let me know.

We wish you well for your research.

Yours sincerely



Nicola Surtees
Chair
Educational Research Human Ethics Committee

"Please note that Ethical Approval and/or Clearance relates only to the ethical elements of the relationship between the researcher, research participants and other stakeholders. The granting of approval or clearance by the Ethical Clearance Committee should not be interpreted as comment on the methodology, legality, value or any other matters relating to this research."

HUMAN ETHICS COMMITTEE

Secretary, Rebecca Robinson
Telephone: +64 03 364 2987, Extn 45588
Email: human-ethics@canterbury.ac.nz

Ref: 2016/13/ERHEC

1 June 2016

Caitlin Duncan
Department of Computer Science and Software Engineering
UNIVERSITY OF CANTERBURY

Dear Caitlin

Thank you for providing the revised documents in support of your application to the Educational Research Human Ethics Committee. I am very pleased to inform you that your research proposal "Teaching Computational Thinking Skills in New Zealand Primary Schools." has been granted ethical approval.

Please note that this approval is subject to the incorporation of the amendments you have provided in your email of 27th May 2016.

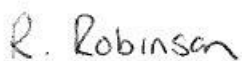
Should circumstances relevant to this current application change you are required to reapply for ethical approval.

If you have any questions regarding this approval, please let me know.

We wish you well for your research.

Yours sincerely

pp



Patrick Shepherd
Chair
Educational Research Human Ethics Committee

Please note that ethical approval relates only to the ethical elements of the relationship between the researcher, research participants and other stakeholders. The granting of approval by the Educational Research Human Ethics Committee should not be interpreted as comment on the methodology, legality, value or any other matters relating to this research.

F E S

HUMAN ETHICS COMMITTEE

Secretary, Rebecca Robinson
Telephone: +64 03 369 4588, Extn 94588
Email: human-ethics@canterbury.ac.nz

Ref: 2016/13/ERHEC Amendment 1

14 June 2017

Caitlin Duncan
Department of Computer Science and Software Engineering
UNIVERSITY OF CANTERBURY

Dear Caitlin

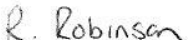
Thank you for your request for an amendment to your research proposal “Teaching Computational Thinking Skills in New Zealand Primary Schools.” as outlined in your email dated 7th June 2017. I am pleased to advise that this amendment has been considered and approved by the Educational Research Human Ethics Committee.

Please note that should circumstances relevant to this current application change you are required to reapply for ethical approval.

If you have any questions regarding this approval, please advise.

We wish you well for your continuing research.

Yours sincerely

PP 

Patrick Shepherd
Chair
Educational Research Human Ethics Committee

Please note that ethical approval relates only to the ethical elements of the relationship between the researcher, research participants and other stakeholders. The granting of approval by the Educational Research Human Ethics Committee should not be interpreted as comment on the methodology, legality, value or any other matters relating to this research.

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